Mechanism Design with Multidimensional, Continuous Types and Interdependent Valuations

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Abstract

We consider the mechanism design problem when agents’ types are multidimensional and continuous, and their valuations are interdependent. If there are at least three agents whose types satisfy a weak correlation condition, then for any decision rule there exist balanced transfers that render truthful revelation a Bayesian ε-equilibrium. A slightly stronger correlation condition ensures balanced transfers exist that induce a Bayesian Nash equilibrium in which agents’ strategies are nearly truthful.

Keywords: Mechanism Design, Interdependent Valuations, Multidimensional Types.

JEL Classification: C70, D60, D70, D82.
1 Introduction

Eliciting private information to guide social decisions is a classic problem of economic theory. For the private-values case, the pioneering work of Vickrey (1961), Clarke (1971), and Groves (1973) shows that if each agent’s preferences depend only on his own information and if the budget need not balance, externality payments make honest revelation a dominant strategy. However, Green and Laffont (1977, 1979) show that dominant strategy implementation is generally incompatible with the requirement that transfers balance the budget. If the solution concept is weakened, positive results are possible. For example, d’Aspremont and Gérard-Varet (1979, 1982) show in the private-values environment that if agents’ beliefs about other agents’ types satisfy a certain condition, which they call compatibility, then for any efficient decision rule there exist balanced Bayesian incentive-compatible transfers that implement it. Later, d’Aspremont, Crémer, and Gérard-Varet (1990) show that when there are three or more agents, the compatibility condition is generically true, and hence that for generic distributions of agents’ types there exists a Bayesian incentive-compatible Pareto-optimal mechanism.

The mechanism design problem has proved more challenging in the case of interdependent valuations, i.e., when one agent’s private information affects other agents’ preferences. Dasgupta and Maskin (2000) study auctions with interdependent valuations and show that a generalized Vickrey auction is efficient if bidders’ types are one dimensional and satisfy a single-crossing property. In general mechanism-design problems, positive results have been mostly limited to the case where agents’ types take on only finitely many values. Work in this area includes Johnson, Pratt, and Zeckhauser (1990); Matsushima (1990; 1991); and McLean and Postlewaite (2004). Crémer and McLean (1985; 1988) study the related question of when it is possible for the designer to earn as much profit as if he were able to observe the agents’ realized private information, the so-called full surplus extraction problem, and show that full extraction is possible when agents’ types are suitably correlated. Aoyagi (1998) considers a model with finite type sets and interdependent valuations and shows that if the distribution of agents’ types satisfies a dependence condition similar to ours, then for any decision rule there exists a balanced, Bayesian incentive-compatible mechanism that

\[1\] Unlike its use in implementation theory (see Jackson, 2001), throughout the paper we use “implement” to refer to the case where there is an outcome of the game that agrees with the decision rule.
implements it.

When types are multidimensional and continuous and valuations are interdependent, the problem becomes even more difficult. After their possibility result for the one-dimensional case, Dasgupta and Maskin (2000) go on to show that when bidders’ types are multidimensional and independently distributed there may be no efficient auction. In a general mechanism design framework, Jehiel and Moldovanu (2001, henceforth JM) explore the difficulties of Bayesian incentive-compatible (BIC) implementation of efficient decision rules when types are multidimensional and continuous and valuations are interdependent. They show that when agents’ types are independently distributed, efficient BIC design is possible only when a certain “congruence condition relating the social and private rates of information substitution is satisfied” (JM, p. 1237). In effect, this congruence condition requires that there be one agent whose relative preference over any two alternatives remains constant for all values of that agent’s information that make the social planner indifferent between those alternatives. They then show that when types are multidimensional the set of payoff functions that satisfy this condition is non-generic, implying that efficient BIC design is generally impossible.²

The present paper addresses the mechanism design problem in environments in which agents’ private information is continuous, multidimensional, and mutually payoff-relevant (i.e., valuations are interdependent). However, we relax the JM assumption that agents’ private information is independently distributed. Our primary interest is to show that when there are three or more agents and agents’ types are stochastically dependent it is possible to design a system of budget-balanced transfer payments that induces agents to (nearly) truthfully reveal their private information and that (nearly) implements any decision rule. In our first result (Theorem 1), we show that under a mild dependence condition on the distribution of agents’ types, which we call Stochastic Relevance, there exist budget-balanced transfers such that for any ε > 0 truthful revelation is an ε-best response to other agent’s truthful announcements, and thus that telling the truth is a Bayesian ε-equilibrium. In our second result (Theorem 2), we show that a slightly stronger version of Stochastic Relevance, which we call Uniform Stochastic Relevance, ensures that there are balanced transfers under which,

²Although Jehiel and Moldovanu (2001) focuses on the impossibility of efficient BIC design, much of the importance of the result lies in the fact that it implies that robust mechanism design using belief-free concepts such as ex post equilibrium is also impossible. We return to this point in Section 5.
for any $\delta > 0$, there is a Bayesian Nash equilibrium (BNE) of the announcement game in which the distance between agents’ equilibrium announcements and their true types is no more than $\delta$, i.e., that there is a nearly truthful BNE. Thus our results provide a complement to those of JM. When the distribution of agents’ types satisfies our dependence assumptions, then incentive-compatible design is possible. Further, our implementation results place very few additional requirements on agents’ preferences.3 In particular, we do not require a single-crossing property.

Mezzetti (2004) considers implementation of efficient decision rules in a model in which the social planner bases transfers on agents’ reports of both their types and the utility they realize from the social decision. The paper shows that implementation of efficient decision rules is generally possible using a two-stage Groves mechanism. However, since agents may not realize the utility from a social decision until long after the decision is made, this framework presupposes, among other things, that the planner is able to make long-term commitments to make transfers in the future. Even in circumstances where the two-stage mechanism is feasible, Mezzetti’s results apply only to efficient decision rules, whereas the results of this paper apply to all decision rules. Further, this present paper imposes a more stringent form of budget balance than Mezzetti.4

McAfee and Reny (1992, henceforth MR) consider the full surplus extraction problem in the case of continuous, multidimensional, and mutually payoff-relevant types with stochastically dependent information. Taking the game played by the agents as given, they show that it is possible to construct for each agent a finite menu of participation fee schedules that extracts almost all of the agent’s rent from playing the game. However, they do not directly address the issue of which decision rules can be implemented, the primary concern of this paper. For example, with multidimensional types and interdependent values, there is, in general, no ex post efficient auction mechanism unless additional assumptions are made that ensure that the agents’ multidimensional information can be summarized by a one-dimensional type (Maskin (1992), Dasgupta and Maskin (2000), Krishna (2002)). Therefore, in such environments, the MR mechanism cannot extract the full information rent (i.e., the rents that would be generated if the auctioneer knew the agents’

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3 Specifically, we require only that agents’ direct returns from the center’s decision be bounded and suitably smooth.
4 We adopt the standard definition in the literature that balanced transfers must sum to zero for any possible choice of actions by the agents. Mezzetti’s transfers satisfy the weaker requirement that the transfers sum to zero on the equilibrium path when all players play truthful strategies.
types), since the MR construction depends on the existence of an ex post efficient mechanism to which the participation fees can be appended. The present paper fills this gap by showing how to construct an ex post efficient mechanism in this environment. Appending the MR mechanism to ours then makes it possible to fully extract the agents’ surplus.

Positive results on incentive-compatible implementation as well as full surplus extraction (e.g., Crémer and McLean (1985, 1988), MR, Aoyagi (1998)) rely on constructing a menu of lotteries for each agent such that the agent maximizes his expected utility when he chooses the lottery intended for his type. Intuitively, this is possible whenever learning an agent’s type provides information about the distribution of the other agents’ types. Our analysis follows in the same spirit.

We capitalize on the literature in statistical decision theory on strictly proper scoring rules, which considers how an informed expert can be induced to truthfully reveal his beliefs about the distribution of future random events. A scoring rule assigns payoffs to the expert based on his announced probabilities for various future events and the event that actually occurs. A strictly proper scoring rule has the property that the decision maker maximizes his expected score when he truthfully announces his beliefs about the distribution. Our incentive-compatibility results rely on payments based on a proper scoring rule to drive agents toward truthful revelation of their private information.

The paper proceeds as follows. Section 2 presents the model. Section 3 constructs scoring-rule payments that render truthful reporting a Bayesian $\varepsilon$-equilibrium. The basic construction is adapted in Section 4 to show that under a slightly stronger correlation condition similarly constructed payments ensure that there is an exact BNE in which agents’ strategies are arbitrarily close to truthful. Section 5 discusses limitations of the approach in the paper, and Section 6 concludes. All proofs are presented in the Appendix.

2 The Model

Suppose $N \geq 3$ agents, indexed by $i = 1, ..., N$, interact with the mechanism designer, whom we will call the center. The center’s task is to elicit agents’ private information in order to choose an

\footnote{See Cooke (1991) and the references therein for a discussion of scoring rules and their uses.}

\footnote{Johnson, Pratt, and Zeckhauser (1992) employs a similar technique in the case of finite type and action spaces.}
alternative $g$ from a set of alternatives $G$.

Each agent $i$ has private information or type $t_i \in T_i$. Agent $i$'s type space $T_i$ is a non-empty, compact, convex subset of $d^i$-dimensional Euclidean space. For each $i$, $d^i$ is a positive integer, and $d^i$ may be different for different agents. We use $T = \times T_i$ to denote the product space of the $N$ agents' type spaces. Following the standard notation we use $t = (t_1, \ldots, t_N)$ for the vector of types, $t_{-i}$ for the vector of all but agent $i$'s type, and $t_{-ij}$ for all but the types of agents $i$ and $j$.

Each agent's utility is quasilinear in his direct return from the social alternative, $g$, and money, $x$, taking the form: $u_i(t, g, x) = V_i(t, g) + x$. Note that agent $i$’s direct return from alternative $g$ depends on all agents’ types. Hence valuations are interdependent.

A decision rule $g : T \rightarrow G$ maps a type for each agent to a social alternative. For simplicity, we assume that $g(t)$ is single valued. For $g(t)$ that is not single-valued, our implementation result applies to any selection from $g(t)$, and therefore this restriction is without loss of generality. Although we will impose a degree of smoothness on $g(t)$, we will not otherwise restrict it. In particular, we do not require that $g(t)$ be efficient.

We consider direct mechanisms in which each agent sends a message (announcement) to the center consisting of an element from his type space. We denote these announcements by $a_i \in T_i$, and let $a$, $a_{-i}$, and $a_{-ij}$ refer respectively to the full announcement vector, the announcement vector leaving off agent $i$, and the announcement vector leaving off agents $i$ and $j$. The remainder of the mechanism consists of a transfer function $x_i(a)$ for each $i$ and a decision rule $g(a)$, with the standard interpretation that the agents announce $a$, social alternative $g(a)$ is realized and transfer $x_i(a)$ is made to agent $i$.

An announcement strategy for agent $i$ is a function $s_i(\cdot) : T_i \rightarrow T_i$ that specifies agent $i$’s announcement in the message game as a function of his information. We will use the notation $s_i(\cdot)$ to refer to a strategy for agent $i$ and $s_i(t_i)$ to denote to the announcement agent $i$ makes under strategy $s_i(\cdot)$ when his type is $t_i$. Thus, $s_i(\cdot)$ is an element of a function space, while $s_i(t_i)$ resides in $d_i$-dimensional Euclidean space. We will use $\tau_i$ to denote the identity function on $\mathbb{R}^{d_i}$, i.e., agent $i$’s truthful strategy.

Denote the vector of transfer rules to all agents by $x(a)$, which we call a transfer scheme. A transfer scheme is balanced if its transfers sum to zero for all possible announcement vectors:
\[ \sum_{i=1}^{N} x_i(a) = 0 \text{ for all } a. \] If a decision rule is implemented by a balanced transfer scheme, it requires no outside subsidy.

Since our mechanism is essentially the same for any decision rule and depends on the decision rule only through the direct return function, we integrate the decision rule into the direct return function, and then write \( V_i(t, g(a)) \) as \( v_i(t, a) \). If there exists a transfer scheme that satisfies a particular solution concept with payoffs \( v_i(t, a) \), then those transfers implement \( g(a) \) under that solution concept. We make the following assumption regarding \( v_i(t, a) \):

**Assumption 1 (Smooth Direct Returns):** *For each \( i \), expected direct returns are twice continuously differentiable in \( a_i \) and \( t_i \).*

Assumption 1 is not innocuous since it implies restrictions on the continuity of the underlying decision rule, \( g(a) \), and on the set of possible decisions, \( G \). Nevertheless, continuity seems to be a reasonable restriction in any situation that is appropriately modeled using continuous types. Further, discontinuous decision rules can often be approximated by continuous ones, and the results below would generalize to the case of decision rules that can be approximated by continuous rules.

Since \( v_i(t, a) \) is continuous and \( T \) is compact, Assumption 1 implies that direct returns are bounded. Let \( \bar{M} \geq 0 \) denote the bound. That is, for any \( i, t_i \) and \( a, |E\{v_i(t, a) | t_i\}| \leq \bar{M} \).

Types are distributed according to commonly known prior distribution \( F(t) \). Let \( f(t_j|t_i) \) be the density of agent \( j \)'s private information conditional on agent \( i \)'s private information, \( t_i \), and let \( f(t_{-i}|t_i) \) be the density of all other agents' private information conditional on agent \( i \)'s private information. We impose two assumptions on agents' beliefs, a smoothness condition and a correlation condition.

**Assumption 2 (Smooth Conditional Distributions):** *For each \( i \) and \( j \neq i \), conditional densities \( f(t_j|t_i) \) and \( f(t_{-i}|t_i) \) are continuous in \( t_j \) and \( t_{-i} \), respectively, and twice continuously differentiable in \( t_i \).*

We assume that the agents’ private information is not independently distributed, which departs from the JM model. Specifically, our informativeness assumption, which we call **Stochastic Relevance**, is that the conditional distribution of the center’s information be different for different values of each agent’s private information.
Assumption 3 (Stochastic Relevance): For each $i$, there exists an agent $j \neq i$ such that for any distinct types $t_i$ and $t'_i$ there exists $t_j$ such that:

$$f(t_j|t_i) \neq f(t_j|t'_i).$$

Let $\|\cdot\|_R$ denote the Euclidean norm, $\|t\|_R = \left(\sum_k (t_{ik})^2\right)^{1/2}$, and $\|\cdot\|_2$ denote the $L_2$ norm, $\|f\|_2 = \left(\int |f|^2 \, ds\right)^{1/2}$. We will write $f_j(\cdot|t_i)$ when we wish to denote agent $i$’s beliefs about the distribution of $t_j$ considered as a function. Lemma 1 follows as an immediate consequence of Assumptions 2 and 3.

Lemma 1: Assumptions 2 and 3 imply that for each $i$, for any $\delta > 0$ there exists $\mu > 0$ such that:

$$\|t_i - t'_i\|_R \geq \delta \implies \|f_j(\cdot|t_i) - f_j(\cdot|t'_i)\|_2 \geq \mu.$$  

Taken together, Assumptions 2 and 3 and Lemma 1 imply that $f(\cdot|t_i)$ and $f(\cdot|t'_i)$ differ on an open subset of $T_j$ and that $f_j(\cdot|t_i)$ and $f_j(\cdot|t'_i)$ are close together (as functions in $L_2$) if and only if $t_i$ is close to $t'_i$. Thus they capture the idea that types should have similar beliefs if and only if they are close together.7

3 Existence of Nearly Bayesian Incentive Compatible Transfers

We begin by considering the question of whether there exist transfers that make the truth nearly a best response, provided that all other agents announce truthfully. Considering this question allows us to illustrate our construction in the simplest setting. In the next section, we go on to show that a similarly constructed payments establish that there is a nearly truthful BNE of the game.

7 Although it would add significant notational burden, Stochastic Relevance could be relaxed to allow for the case where agent $i$’s beliefs about the joint distribution of a group of agents’ types depends on $t_i$ even though the marginal distribution for any other agent’s type does not. Aoyagi (1998) presents such a condition (Assumption 2) for the finite case.
We begin with the notion of $\varepsilon$-Bayesian Incentive Compatibility. Transfer scheme $x_i(a)$ is $\varepsilon$-Bayesian Incentive Compatible ($\varepsilon$-BIC) if for any $i$, $t_i$, and $a_i$:

$$E \{v_i(t, t_{-i}, t_i) + x_i(t_{-i}, t_i) | t_i\} \geq E \{v_i(t, t_{-i}, a_i) + x_i(t_{-i}, a_i) | t_i\} - \varepsilon.$$ 

That is, if for each agent $i$, announcing truthfully is an $\varepsilon$-best response to the other agents’ truthful announcements.

As discussed earlier, the mechanism we propose draws on the decision-theoretic literature on proper scoring rules. In particular, we employ the quadratic scoring rule. Suppose that agent $j$ is using the truthful announcement strategy, $s_j(\cdot) = \tau_j$, and player $i$ is being scored based on how well he predicts agent $j$’s announced type. The quadratic score assigned to type $t_j$ when agent $i$ announces $a_i$ is given by:

$$Q(t_j|a_i) = 2f(t_j|a_i) - \int_{T_j} f(t_j|t_i)^2 dt_j.$$ 

Lemmas 2 and 3 establish basic properties of the quadratic scoring rule that will be used in the subsequent analysis.

**Lemma 2:** For agent $i$, choose an agent $j$ according to Assumption 3, and suppose agent $j$ truthfully announces his type, $s_j(\cdot) = \tau_j$. Truthful revelation uniquely maximizes agent $i$’s expected quadratic score:

$$t_i = \arg \max_{a_i \in T_i} \int_{T_j} Q(t_j|a_i) f(t_j|t_i) dt_j.$$ 

As Selten (1988) notes, the proof that truthful revelation uniquely maximizes the expected quadratic score also shows that the expected loss from agent $i$’s announcing $a_i \neq t_i$ instead of his true type $t_i$ is equal to the square of the $L_2$-distance between agent $i$’s beliefs when his type is $a_i$ and when his type is $t_i$. Lemma 3 exploits this property.

**Lemma 3:** For agent $i$, choose an agent $j$ according to Assumption 3. For any $\delta > 0$ there exists $\varepsilon$-Bayesian Incentive compatibility appears, for example, in d’Aspremont and Gérard-Varet (1982).
\[ \varepsilon > 0 \text{ such that the expected quadratic score for the distribution of agent } j \text{'s type from announcing } a_i \neq t_i \text{ with } \|a_i - t_i\|_R > \delta \text{ is at least } \varepsilon \text{ worse than announcing truthfully:} \]

\[
\|a_i - t_i\|_R \geq \delta \text{ implies } \int_{T_j} Q(t_j|a_i) f(t_j|t_i) dt_j - \int_{T_j} Q(t_j|a_i) f(t_j|a_i) dt_j \geq \varepsilon.
\]

Lemma 2 establishes that if the agents care only about the transfer, truthful announcement is agent \( i \)'s unique best response when other agents' tell the truth. Lemma 3 ensures that there is no sequence of announcements far away from the truth whose expected scores converge to the expected score of the truth. This is needed in order to establish a uniform lower bound on the loss from an announcement that is far from truthful.

Our first main results shows that there exist \( \varepsilon \)-BIC, balanced transfers. The intuition is that in choosing whether to announce his true type or some other type the agent weighs the effects of lying on the expected transfer and on the expected direct return. If transfers are based on the quadratic scoring rule, then telling the truth maximizes the agent’s expected transfer. However, since announcing truthfully does not necessarily maximize the expected direct return, the agent may have an incentive to deviate from truth-telling, sacrificing expected transfer in order to enjoy a personally superior social alternative. Of course, the agent’s willingness to do so depends on how quickly the expected transfer declines relative to the increase in expected direct return. By scaling up the scoring-rule based payments to the agent, the center can increase the importance of the transfer loss relative to the direct return gain, making anything but a small deviation from the truth unprofitable.

**Theorem 1:** Under Assumptions 1 - 3, for any decision rule and any \( \varepsilon > 0 \) there exist \( \varepsilon \)-BIC, balanced transfers.

The essence of the proof is to divide agent \( i \)'s announcements into two groups – those that are within \( \delta \) of the truth and those that are not. Under the quadratic scoring rule, the expected transfer is maximized by telling the truth. Thus announcements that are within \( \delta \) of the truth yield a smaller expected transfer but a possibly larger direct return. However, by choosing \( \delta \) sufficiently
small we ensure that the direct return gain from any announcement within $\delta$ of the truth must be less than $\varepsilon$. On the other hand, Assumptions 2 and 3 ensure that the loss in expected transfer from moving from a truthful announcement to one that is more than $\delta$ from the truth must be uniformly bounded away from zero, and thus scaling up the transfers increases the minimum loss in transfer from an announcement at least $\delta$ from the truth. Since direct returns are bounded, a sufficient scaling of the transfers ensures that the gain direct return gain cannot outweigh the transfer loss, and thus that announcements that are at least $\delta$ from the truth must involve a total expected utility loss of at least $\varepsilon$.

The transfers are balanced using a permutation construction. That is, if agent 1 is given incentives to report truthfully by comparing his announcement to that of agent 2, then the transfer to agent 1 can be funded by a third agent (e.g., agent 3) without affecting any agent’s incentive to report truthfully. Repeating this process for all agents balances the budget. Thus, while three or more agents are needed in order to balance the budget, if budget balance is not a concern, $\varepsilon$-BIC transfers exist with only two agents.

4 Existence of a Nearly Truthful Bayesian Nash Equilibrium

Theorem 1 establishes that compensating agents using a sufficiently large scaling of the quadratic scoring rule renders truthful revelation an $\varepsilon$-best response, provided that the other agents announce truthfully. Although this idea has some intuitive appeal and makes the role of the quadratic scoring rule transparent, the introduction of satisficing behavior is somewhat arbitrary, and it begs the question of whether this limited rationality is necessary or merely a convenience. To address this concern, we next argue that, under reasonable conditions, payments based on a scaling of the quadratic scoring rule can be used to induce a BNE in which agents’ strategies are arbitrarily close to the truth.

For a fixed transfer scheme $x(a)$, a BNE of the announcement game is a vector of strategies $(s_1(\cdot),..., s_N(\cdot))$ such that for each $i$ and $t_i$: \[
s_i(t_i) \in \arg \max_{a_i} E_{t_{-i}} \left\{ v_i(t_i, a_i, s_{-i}(\cdot)) + x_i(a_i, s_{-i}(\cdot) | t_i) \right\}.
\]
We endow the space of announcement strategies with the sup norm:

$$\|s_i(\cdot) - \hat{s}_i(\cdot)\|_{sup} = \sup_{t_i} \left( \sum_{n=1}^{d_i} (s_{in}(t_i) - \hat{s}_{in}(t_i))^2 \right)^{1/2}.$$ 

For $\delta > 0$, we call an announcement strategy, $s_i(\cdot)$, $\delta$-truthful if $\|s_i(\cdot) - \tau_i\|_{sup} \leq \delta$. That is, a $\delta$-truthful announcement strategy is one in the agent’s announcement is always within distance $\delta$ of his true type. We say that transfer scheme $\delta$-implies a decision rule in BNE if under those transfers there exists a BNE in which all agents’ strategies are $\delta$-truthful.\footnote{We may, on occasion, refer to single announcements as $\delta$-truthful if for a particular $t_i$, $\|s_i(t_i) - t_i\|_R \leq \delta$ or to strategy profiles as being $\delta$-truthful if each individual strategy is $\delta$-truthful.} Note that the concept of $\delta$-implementation in BNE allows for the existence of BNE that are not $\delta$-truthful.

Let $C_i$ denote the space of continuous announcement strategies for agent $i$. For $\delta > 0$, let $C_i(\delta)$ the space of continuous, $\delta$-truthful announcement strategies for agent $i$: $C_i(\delta_i) \equiv \left\{ s_i(\cdot) \in C_i : \|s_i(\cdot) - \tau_i\|_{sup} \leq \delta \right\}$. Let $C(\delta)$ be the product space $\times C_i(\delta)$, and $C_{-i}(\delta)$ be defined in the usual way as the product space of $C_j(\delta)$ for all agents except $i$, each endowed with the appropriate product topology.

The key step in constructing a $\delta$-truthful BNE is ensuring that a version of Stochastic Relevance remains true even when agents’ announcements are only $\delta$-truthful. In order to ensure this we strengthen stochastic relevance as follows:

**Assumption 4 (Uniform Stochastic Relevance):** There exists $\phi > 0$ such that for each $i$, there exists an agent $j \neq i$ such that for any distinct types $t_i$ and $t'_i$ there exists an open ball $\theta_j(t_i, t'_i) \subset T_j$ with radius $\phi$ such that $f(t_j|t_i) \neq f(t_j|t'_i)$ for all $t_j \in \theta_j(t_i, t'_i)$.

Stochastic Relevance (Assumption 3) implies that, for any distinct types $t_i$ and $t'_i$, $f(t_j|t_i)$ and $f(t_j|t'_i)$ differ on an open set of types for agent $j$. Uniform Stochastic Relevance (Assumption 4) strengthens Stochastic Relevance by requiring that there be a lower bound on the size of the open set on which $f(t_j|t_i)$ and $f(t_j|t'_i)$ differ that is independent of the particular pair of types $t_i$ and $t'_i$ that is chosen. It is straightforward to show that, by virtue of compactness, Assumption 3 implies the existence of such a uniform bound provided that $t_i$ and $t'_i$ are bounded away from each other, i.e., as long as there exits $\delta > 0$ such that $\|t_i - t'_i\|_R \geq \delta$. Thus, to the extent that Uniform
Stochastic Relevance is stronger than Stochastic Relevance, it only restricts the behavior of beliefs as types $t_i$ and $t'_i$ become (arbitrarily) close together.

Since agent $i$’s beliefs are continuous in $t_i$, as $t_i$ and $t'_i$ become very close, the two types’ beliefs must also become very close. Uniform Stochastic Relevance rules out the case in which as $t'_i$ converges $t_i$ the Lebesgue measure of the set of $t_j$ where their associated beliefs differ, $\{t_j \in T_j | f(t_j|t_i) \neq f(t_j|t_i')\}$, converges to zero. In other words, under Uniform Stochastic Relevance it cannot be that as $t'_i$ approaches $t_i$, $f(t_j|t'_i)$ approaches $f(t_j|t_i)$ by becoming equal to it on an ever-larger set of $t_j$. Seen in this way, it is clear that many of the natural families of beliefs will satisfy Uniform Stochastic Relevance.

For an example of a family of beliefs that does not satisfy Uniform Stochastic Relevance, consider $T_i = T_j = [0,1]^2$. Suppose that $f(t_j|t_i)$ is uniformly distributed on a disk centered at $t_j = t_i$ and having radius $1/10$ (for $t_i$ suitably distant from the boundary of $T_j$). Consider $t_i = (1/2,1/2)$. Let $\lambda(\cdot)$ denote Lebesgue measure. Since

$$\lim_{\|t'_i - (1/2,1/2)\| \to 0} \lambda \left( \left\{ t_j \in T_j | f(t_j|t_i) \neq f\left(t_j|t'_i\right) \right\} \right) = 0,$$

these beliefs violate Uniform Stochastic Relevance. This is because $f(t_j|t_i)$ and $f(t_j|t'_i)$ are equal on their common support, and as $t'_i$ converges to $t_i$, the supports of $f(t_j|t_i)$ and $f(t_j|t'_i)$ converge as well.\(^{10}\)

The existence of a uniform lower bound on how often the beliefs of two different types of agent $i$ differ is important because Theorem 2 considers $\delta$-truthful strategies. If $f(t_j|t_i)$ and $f(t_j|t'_i)$ differ, the distribution of agent $j$’s announcements resulting from a particular $\delta$-truthful strategy (e.g., the $\delta$-truthful strategy that is constant over the set of $t_j$ where $f(t_j|t_i)$ and $f(t_j|t'_i)$ differ) is the same for $t_i$ and $t'_i$. Lemma 4 shows that Uniform Stochastic Relevance ensures that different types $t_i$ and $t'_i$ have different beliefs about the distribution over a set of discrete events comprised of groups of announcements for some agent $j \neq i$, and that this difference remains even if agent $j$.

\(^{10}\)On the other hand, if $f(t_j|t_i)$ is distributed as a cone with a circular base of radius $1/10$ and peak at $t_i$, these beliefs would satisfy Uniform Stochastic Relevance since the set of points where the densities of $f(t_j|t_i)$ and $f(t_j|t'_i)$ are equal remains small (i.e., has Lebesgue measure zero) even as $t_i$ and $t'_i$ become arbitrarily close together.
distorts his announcement slightly.\textsuperscript{11,12}

**Lemma 4:** Assumptions 2 and 4 imply that there exists $\delta^* > 0$ such that for any $0 < \delta < \delta^*$ and any agent $i$, there is an agent $j \neq i$ such that $T_j$ contains a finite set of disjoint balls $B^{ij} = \{b^{ij}_1, \ldots, b^{ij}_M\}$ with radius at least $\delta$ such that for any $t_i, t'_i$ with $t_i \neq t'_i$ there is at least one $b^{ij}_M$ such that $f(t_j | t_i) \neq f(t_j | t'_i)$ for all $t_j \in b^{ij}_M$.

To see the role that Lemma 4 will play in the proof of Theorem 2, consider two distinct types $t_i$ and $t'_i$ for agent $i$. By Lemma 4, let $b^{ij}_m$ be the ball in agent $j$’s announcement space satisfying Lemma 4 that distinguishes these types. If agent $j$ announces truthfully, types $t_i$ and $t'_i$ assign different probabilities to event $t_j \in b^{ij}_m$, and so a scoring rule based on whether agent $j$’s announcement is in $b^{ij}_m$ can be used to truthfully elicit whether agent $i$’s type is $t_i$ or $t'_i$.

The lower bound on the size of the balls in $B^{ij}$ ensures that there is a partition of events that distinguishes any two types even if $j$ is allowed to distort his announcements using a $\delta$-truthful strategy (for $\delta < \delta^*$). To see how, let $b^{ij}_m$ be the ball in $B^{ij}$ to which $t_i$ and $t'_i$ assign different probabilities to the event $t_j \in b^{ij}_m$. Let $r_m$ be the radius of $b^{ij}_m$, and let $\hat{b}^{ij}_m$ be the ball with the same center and radius $r_m - \delta$.

If $j$’s strategy is $\delta$-truthful, then the set of types that announces $a_j \in \hat{b}^{ij}_m$ must be contained in $b^{ij}_m$. Hence whenever $s_j(\cdot) \in C_j(\delta)$, types $t_i$ and $t'_i$, assign different probabilities to the event $a_j \in \hat{b}^{ij}_m$ conditional on $s_j(\cdot)$. To see why, consider Figure 1, which illustrates the one-dimensional case. Suppose that $\hat{b}^{ij}_m = [t^+_j - \delta, t^+_j + \delta]$. According to Lemma 4, the densities for some $t_i$’s are drawn in such that they don’t cross over this region. Now, look at the smaller event, $\hat{b}^{ij}_m = [t^-_j, t'^+_j] \subset b^{ij}_m$.

Note that since types can only distort their announcements by $\delta$ or less, if $t_j$ $\delta$-truthfully announces $a_j \in \hat{b}^{ij}_m$, then $t_j \in b^{ij}_m$. However, since the densities for these values of $t_i$ are ranked over the entire set $\hat{b}^{ij}_m$, conditional on $s_j(\cdot) \in C_j(\delta)$, two distinct types whose densities do not cross over $b^{ij}_m$ cannot assign the same probability to $j$’s announcement being in $\hat{b}^{ij}_m$. In Figure 1, the heavy black lines on

\textsuperscript{11}Since Uniform Stochastic Relevance implies Lemma 4, it is a stronger condition. However, it is also more straightforward to verify than Lemma 4. Alternatively, we could have assumed the weaker condition (Lemma 4) directly.

\textsuperscript{12}Lemma 4 is also useful for a more technical reason. When $s_j(\cdot) \in C_j(\delta)$, player $j$’s announcement can be constant over an open interval. Hence, even though $t_j$ has a density, the distribution of $j$’s announcements can have point masses. While virtually the same theory of proper scoring rules applies either to discrete or continuous distributions, to our knowledge there is no theory of proper scoring rules for mixed distributions. Therefore we move to using a scoring rule for the distribution over discrete events.
The horizontal axis indicate the set of types $t_j$ that make announcements in $\hat{b}_{ij}^m$ for some hypothetical $s_j(\cdot) \in C_j(\delta)$. Looking at the shaded regions above the $t_j$ in this set, the densities for the various values of $t_i$ do not cross. Thus for any two $t_i$ whose densities do not cross over $b_{ij}$, the one with the higher density must assign higher probability to the event $a_j \in \hat{b}_{ij}^m$ for any possible announcement strategy $s_j(\cdot) \in C_j(\delta)$.

If, as asserted by Uniform Stochastic Relevance and Lemma 4, there is a finite set of balls $B_{ij}$ such that for each possible pair of types there is one ball over which their densities do not cross, then we can use the sets $\hat{b}_{ij}^1, \ldots, \hat{b}_{ij}^M$ along with $\hat{b}_{ij}^0 \equiv T_j \setminus \hat{b}_{ij}^m$ (i.e., “everything else”) as a partition of events that distinguishes every pair of types for every possible strategy $s_j(\cdot) \in C_j(\delta)$. That is, conditional on $s_j(\cdot)$, different types $t_i$ have different beliefs about the distribution over events in $\hat{B}_{ij} = \{\hat{b}_{ij}^0, \hat{b}_{ij}^1, \ldots, \hat{b}_{ij}^M\}$. Thus, transfers based on the quadratic scoring rule applied to the events in $\hat{B}_{ij}$ (conditional on $s_j(\cdot)$) are strictly proper. If agent $i$ only cared about the transfer, his best response to $s_j(\cdot)$ under these transfers would be to announce his true type. When agent $i$ also cares about the direct return from the social choice, basing transfers on a sufficiently large scaling of the quadratic scoring rule ensure that agent $i$’s best response is close to truthful.

Theorem 2 establishes that there is a transfer scheme that $\delta$-implements any decision rule in BNE. The main tool employed is Schauder’s fixed point theorem (see Zeidler, 1985), and our proof.
draws heavily on Meirowitz’s (2003) general existence result for equilibria in Bayesian games with infinite type and action spaces.

**Theorem 2:** *Under assumptions 1, 2, and 4, for any decision rule and any $\delta > 0$ there exist balanced transfers that $\delta$-implement that decision rule in BNE.*

The intuition for the proof begins by noting that each agent’s payoff is a linear combination of his direct return and transfer. Thus, the situation where the transfers are multiplied by a large (positive) constant is one where the agent puts small relative weight on his direct return, which is similar to the case where the agent puts zero weight on his direct return. When agents care only about their transfers, transfers based on the quadratic scoring rule ensure that truthful revelation is a strict equilibrium. If we knew that this equilibrium changed smoothly with the relative weight put on agents’ direct returns, then games with nearby payoffs would have a nearby equilibrium. Thus, games in which the relative weight on transfers was very high would have an equilibrium in which agents’ strategies were nearly truthful. Unfortunately, this smooth dependence property, which is related to lower hemi-continuity of the equilibrium correspondence, does not hold in general. Nevertheless, by exploiting the fact that the truthful equilibrium of the transfers-only game is strict, we are able to show that when agents’ strategies are nearly truthful, nearby games satisfy the requirements for the application of Schauder’s fixed point theorem, and thus that games where agents’ put small relative weight on their direct returns have a nearly truthful equilibrium.\(^{13}\)

Theorem 2 establishes the existence of an equilibrium in which agents play nearly truthful strategies. The question remains whether, under the transfers that induce the $\delta$-truthful equilibrium, other equilibria exist as well and, if so, whether those equilibria are also $\delta$-truthful. In general, there is no reason to rule out such equilibria. Given that agents’ incentives are primarily driven by their desire to maximize the transfer they receive, and that these transfers are determined by how well each agent predicts the announcements of the other agents, it is easy to imagine that there could be equilibria in which all agents permute their announcements in such a way that announcements are no longer close to truthful but still predict other agents’ announcements well.

\(^{13}\) The key step is to establish that agents’ best responses to any $\delta$-truthful opponents’ strategies are unique, which requires conditioning transfers on the other agents’ strategies.
There is, however, one circumstance in which it is possible to establish that all equilibria must be nearly truthful. This is the case in which the center receives a signal of its own that is stochastically related to the agents’ types.\textsuperscript{14} In effect, for each agent $i$, the center’s information plays the role of the agent $j$ whose information is used to score agent $i$. Since no agent’s behavior can affect the distribution of the center’s information, the expected payment from transfers based on the quadratic scoring rule applied to the center’s information is uniquely maximized by telling the truth regardless of the other agents’ strategies. The argument in Theorem 2 then establishes that for any $\delta > 0$ there exists a $\delta$-truthful equilibrium. Further, since payments can be scaled up sufficiently that all best responses are $\delta$-truthful, all equilibria must be $\delta$-truthful.

Returning to the case where the center does not receive an informative signal, while Theorem 2 establishes that agents’ strategies are nearly truthful, from the perspective of the social planner our real interest is not whether agents are telling the truth, but rather whether the resulting social choice rule is close to that implied by the planner’s desired rule, and whether realized social welfare is close to the planner’s desired welfare level. These desirable properties follow, however, because transfers are balanced and agents’ payoffs are assumed to be continuous conditional on the social choice function (Assumption 1).

MR shows that when agents’ types are correlated, for any game the center can extract from each agent nearly all of the rents that agent earns by participating in the game.\textsuperscript{15} Their mechanism offers agents a finite menu of participation fee schedules such that, when the agent selects his preferred schedule and then plays the game, he is left with a rent that, though positive, is arbitrarily small.

While MR shows that for a given game, a participation fee schedule can ensure that agents’ interim participation constraints can be satisfied at (nearly) no cost to the center, they do not address the question of whether, for a given decision rule, a game exists that implements that decision rule. In particular, the center cannot extract the full information rent (i.e., the rent that would be generated if the center could observe agents’ types and make ex post efficient decision) unless there exists a mechanism that implements the ex post efficient decision rule. Prior to this paper, there have been no results that show the general existence in the standard mechanism design

\textsuperscript{14}If the Center’s information has a density, then the center’s information must satisfy the player $j$ role in Assumption 3.

\textsuperscript{15}Their required condition (*) is strictly stronger than our Assumption 3.
framework of an ex post efficient mechanism when agents have multidimensional, continuous types and interdependent valuations.\textsuperscript{16} To the extent that ex post efficient mechanisms have been shown to exist, these results typically require additional assumptions on the form of agents’ direct returns functions, e.g., single crossing. The results in this paper do not impose any restrictions on direct returns functions beyond smoothness (Assumption 1).

We show that if agents’ types are correlated, then any decision rule, including the ex post efficient decision rule, can be implemented arbitrarily closely (i.e., $\delta$-truthfully). Provided that beliefs satisfy MR’s condition (*), our result, coupled with the MR result, establishes that the center can extract (approximately) the full information rent and satisfy agents’ interim participation constraints by first offering agents a menu of participation fees and then running our scoring-rule based system.

5 Limitations of Quasilinear Mechanism Design

This paper employs the quasilinear mechanism design framework, and as such it suffers from the well-known limitations of the approach.\textsuperscript{17} These include, first, that the transfers needed to induce (near) truth-telling may be very large, and thus for small $\delta$ our mechanism may be infeasible if agents face limited liability constraints. Second, the quasilinear framework assumes that agents are risk neutral with regard to the transfers. If agents are risk averse over the transfer, then it will not generally be possible to (nearly) implement any decision rule with budget balance. However, if the center is interested in inducing (nearly) truthful revelation without budget balance, then redefining the transfers in terms of utilities instead of monetary amounts will accomplish this goal.

Recently, Neeman (2004) and Heifetz and Neeman (2006) have launched another line of criticism against the literature on mechanism design with correlated information. They argue that, although the correlation requirements employed in the literature appear rather reasonable, they have the common feature that an agent’s beliefs uniquely determine his preferences, which they term the BDP property. Stochastic relevance, as embodied in Assumptions 3 and 4 of this paper, implies

\textsuperscript{16}As discussed earlier, Mezzetti’s (2004) mechanism operates in a slightly different framework than the standard models and uses a weaker form of budget balance.

\textsuperscript{17}Crémer and McLean (1988) discuss the limited liability and risk neutrality assumptions in the context of their full extraction result.
the BDP property. Heifetz and Neeman (2006) show that the BDP property is a non-generic property of the universal type space. Thus, while correlation seems like a reasonable assumption, the set of BDP beliefs is, in a sense, “small.”

Another line of criticism regarding Bayesian mechanism design is that Bayesian equilibrium is belief-based. As such, incentive compatible mechanisms are highly sensitive to the information structure of the problem. MR observes that such dependence casts doubt on whether such results teach us much about real-world asymmetric information problems. This criticism has led to the search for “robust” mechanisms that do not depend on agents’ beliefs about others’ information, usually employing the concept of ex post equilibrium. In light of this, the JM result on the generic impossibility of efficient BIC design with independently distributed types also implies the impossibility of ex post incentive compatibility. Our results provide a counterpoint to JM by showing that BIC design is possible in their environment if the independence assumption is relaxed. However, our BIC mechanism is not ex post incentive compatible. Indeed, Jehiel et al. (2006) show that only constant decision rules are implementable in ex post equilibrium in generic mechanism design problems with multidimensional, continuous types and interdependent valuations.

6 Conclusion

This paper extends the mechanism design literature to show the possibility of incentive-compatible implementation of any decision rule when agents’ types are continuous, multidimensional, and mutually payoff relevant, provided that agents’ types are suitably correlated. Thus we provide a complement to the JM impossibility result for the case of independent information. Our results also complement MR by showing that there is an ex post efficient mechanism in the multidimensional, continuous, mutually payoff-relevant case.

While we show the existence of transfers that induce a $\delta$-truthful BNE, we have not considered the question of whether there exist transfers that render the exact truth a BNE. This is a technically daunting task that remains an open question.

The scoring-rule based approach we adopt has the advantage of being simpler than those commonly adopted in the mechanism design literature. Stochastic relevance (as embodied in Assump-
tion 3 or 4) requires verifying only that distributions are different for different types, which is substantially easier than verifying the compatibility condition of d’Aspremont and Gérard-Varet (1979; 1982), the linear independence condition of Crémer and McLean (1985; 1988), or the generalization of the Crémer-McLean condition found in MR, each of which must hold for all prior distributions for each agent’s type. In addition to being simpler, stochastic relevance is also slightly weaker than any of these conditions. The scoring-rule-based payments used in our mechanism are also relatively simple to construct and our proofs provide a blueprint for doing so. This is an advance over traditional approaches, which generally prove the existence of a mechanism but provide little guidance as to how to construct it.

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18 To be fair, the task of full surplus extraction is more demanding than (nearly) truthful implementation, and so while these papers employ stricter conditions they also achieve stronger results. Our condition is very similar to that employed by Aoyagi (1998) in the finite case.

19 Frequently, such approaches rely on a linear systems approach to demonstrate existence. See d’Aspremont, Crémer, and Gérard-Varet (1990) for a survey of the use of this method.
References


Appendix: Proofs

Proof of Lemma 1: Suppose not. Then there exists $\delta > 0$ such that for all $n$ there exist $t_{in}$ and $t_{in}'$ such that $||t_{in} - t_{in}'||_R \geq \delta$ and $||f_j (\cdot|t_{in}) - f_j (\cdot|t_{in}')||_2 \leq 1/n$. By compactness, $(t_{in}, t_{in}')$ has a convergent subsequence. Let the limit be $(t_{i\infty}, t_{i\infty}')$, and note that $t_{i\infty} \neq t_{i\infty}'$. We have that $\lim ||f_j (\cdot|t_{in}) - f_j (\cdot|t_{in}')||_2 = 0$, and:

$$
\lim ||f_j (\cdot|t_{in}) - f_j (\cdot|t_{in}')||_2 = \lim \left( \int |f_j (t_j|t_{in}) - f_j (t_j|t_{in}')|^2 dt_j \right)^{1/2} \\
\geq \lim \left( \int \min \left\{ 1, |f_j (t_j|t_{in}) - f_j (t_j|t_{in}')|^2 \right\} dt_j \right)^{1/2} \\
= \left( \int \lim \left\{ \min \left\{ 1, |f_j (t_j|t_{in}) - f_j (t_j|t_{in}')|^2 \right\} \right\} dt_j \right)^{1/2} \\
= \left( \int \min \left\{ 1, \lim |f_j (t_j|t_{in}) - f_j (t_j|t_{in}')|^2 \right\} dt_j \right)^{1/2}
$$

where the fact that $\min \left\{ 1, |f_j (t_j|t_{in}) - f_j (t_j|t_{in}')|^2 \right\} \leq 1$ allows us to apply Lebesgue’s Dominated Convergence Theorem in moving from the second to third line. Since

$$
\left( \int \min \left\{ 1, |f_j (t_j|t_{i\infty}) - f_j (t_j|t_{i\infty}')|^2 \right\} dt_j \right)^{1/2} = 0,
$$

$|f (t_j|t_{i\infty}) - f (t_j|t_{i\infty}')| = 0$ for almost all $t_j$. Since $f (t_j|t_i)$ is continuous, $f (t_j|t_{i\infty}) = f (t_j|t_{i\infty}')$ for all $t_j$. However, this violates Assumption 3.

Proof of Lemma 2: The result is standard. This proof follows Selten (1998). Let $\Upsilon (a_i|t_i) = \int Q (t_j|a_i) f (t_j|t_i) dt_j$ be agent $i$’s expected transfer when $s_j (\cdot) = \tau_j$. Substituting in the definition of the quadratic scoring rule, we have:

$$
\Upsilon (a_i|t_i) = \int_{T_j} \left( 2f (t_j|a_i) - \int_{T_j} f (t_j|a_i)^2 dt_j \right) f (t_j|t_i) dt_j.
$$

Rearranging $\Upsilon (a_i|t_i)$ yields:

$$
\Upsilon (a_i|t_i) = \int_{T_j} f (t_j|t_i)^2 dt_j - \int_{T_j} (f (t_j|a_i) - f (t_j|t_i))^2 dt_j.
$$

Hence:

$$
\Upsilon (a_i|t_i) - \Upsilon (t_i|t_i) = - \int_{T_j} (f (t_j|a_i) - f (t_j|t_i))^2 dt_j,
$$

which is zero when $a_i = t_i$ and strictly negative otherwise.■

Proof of Lemma 3: From Lemma 1, for any $\delta$ there exists a $\mu > 0$ such that
Consider the expected utility reaped by a truthful announcement as compared to announcing $i$ for agent $a_i$ and $t_i$ with $||a_i - t_i|| \geq \delta$. Using the notation from the proof of Lemma 2,

$$\left| \int_{T_j} Q(t_j|a_i) f(t_j|t_i) dt_j - \int_{T_j} Q(t_j|a_i) f(t_j|a_i) dt_j \right|$$

$$= |\mathcal{Y}(a_i|t_i) - \mathcal{Y}(t_i|t_i)|$$

$$= \left| - \int_{T_j} (f(t_j|a_i) - f(t_j|t_i))^2 dt_j \right|$$

$$= (||f(t_j|a_i) - f(t_j|t_i)||_2^2 \geq \mu^2.$$ 

Letting $\varepsilon = \mu^2$ completes the proof. \qed

**Proof of Theorem 1:** Consider agent $i$, and suppose all other agents announce truthfully, $s_j(\cdot) = \tau_j, \forall j \neq i$. Thus for the remainder of the proof we can replace $a_j$ with $t_j$ in the expression for agent $i$’s payoff. Since expected direct returns are continuous in announcements and $\times T_i$ is compact, expected direct returns are uniformly continuous. Hence for any $\varepsilon > 0$ there exists $\delta > 0$ such that $||t_i - a_i||_R \leq \delta$ implies $|E(v_i(t, t_{-i}, a_i)|t_i) - E(v_i(t, t_{-i}, t_i)|t_i)| \leq \varepsilon$.

Given $\delta$, let $T'_i = \{(t_i, a_i) \in T_i \times T_i : ||t_i - a_i||_R \geq \delta\}$ be the set of type-announcement pairs that are at least $\delta$ apart. For each $i$, choose $j(i)$ according to Assumption 2. Define the payments to agent $i$ according to the quadratic scoring rule:

$$x_i(t_{-i}, a_i) = 2f(t_{j(i)}|a_i) - \int_{T_{j(i)}} f(t_{j(i)}|a_i)^2 dt_{j(i)}.$$  \hspace{1cm} (1)

Since the expected quadratic score is uniquely maximized at $a_i = t_i$ and $T'_i$ is compact, by Lemma 3 there exists an $\hat{\varepsilon} > 0$ such that for any $(t_i, a_i) \in T'_i$:

$$E\{x_i(t_{j(i)}, a_i)|t_i\} \leq E\{x_i(t_{j(i)}, t_i)|t_i\} - \hat{\varepsilon}.$$  \hspace{1cm} (2)

Next, scale the payments to agent $i$ according to $x^*_i(t_{-iq(i)}, a_i)$:

$$x^*_i(t_{j(i)}, a_i) = \frac{2M + 1}{\hat{\varepsilon}}x_i(t_{j(i)}, a_i).$$  \hspace{1cm} (3)

Consider the expected utility reaped by a truthful announcement as compared to announcing $a_i$ with $(t_i, a_i) \in T'_i$.

$$E\{v_i(t, t_{-i}, a_i) + x^*_i(t_{j(i)}, a_i)|t_i\} - E\{v_i(t, t_{-i}, t_i) + x^*_i(t_{j(i)}, t_i)|t_i\}$$

$$= E\{v_i(t, t_{-i}, a_i) - v_i(t, t_{-i}, t_i)|t_i\} + E\{x^*_i(t_{j(i)}, a_i) - x^*_i(t_{j(i)}, t_i)|t_i\}$$

$$< 2M + (2M + 1) \left[ \frac{x_i(t_{j(i)}, a_i)}{\hat{\varepsilon}} - \frac{x_i(t_{j(i)}, t_i)}{\hat{\varepsilon}} \right]$$

$$< 2M + (2M + 1) \left[ -\frac{\hat{\varepsilon}}{\hat{\varepsilon}} \right] < 0.$$ 

Hence under payment scheme $x^*_i(t_{j(i)}, a_i)$, announcing truthfully earns a higher payoff than any announcement such that $(a_i, t_i) \in T'_i$. And, by the choice of $\delta$, announcements $a_i$ such that
(a_i, t_i) \notin T'_i \text{ have lower expected transfers than truthful announcements and have expected direct returns that exceed those of truthful announcement by less than } \epsilon. \text{ Hence truthful announcement is an } \epsilon\text{-best response. Since } i \text{ is chosen arbitrarily, payments can be constructed that make the truth an } \epsilon\text{-best response for all agents.}

To balance the budget, for each agent } i \text{ choose an agent } x(i) \notin \{i, j(i)\} \text{ with the understanding that } x(i) \text{ will fund } i\text{'s transfer. Let } K_i = \{k | x(k) = i\} \text{ be the set of all agents whose transfers } i \text{ funds. Agent } i\text{'s net transfer is therefore:

\[
x_i(t_i, a_i) = x^*_i(t_{j(i)}, a_i) - \sum_{k \in K_i} x^*_k(t_{j(k)}, a_k).
\]

Since agent } i\text{'s announcement does not affect the terms after the summation, his incentives are not affected, and transfer scheme } x(t, a) \text{ is } \epsilon\text{-BIC and balances the budget.}

**Proof of Lemma 4:** For each } i \text{ choose an appropriate } j \text{ according to Assumption 4. Overlay a } d_i\text{-dimensional rectangular grid over } T_j \text{ by dividing each of the } d_i \text{ dimensions into increments of size } \beta > 0. \text{ Thus, the grid divides } T_j \text{ into hypercubes with sides of length } \beta. \text{ The maximum distance between any two points in a hypercube is } \beta\sqrt{d_i} \text{ (i.e., the distance in } \mathbb{R}^{d_i} \text{ between } (0,0,...,0) \text{ and } (\beta,\beta,...,\beta)). \text{ Choose } \beta \text{ such that } \beta\sqrt{d_i} < \phi. \text{ Consider two distinct types } t_i \text{ and } t'_i \text{ for agent } i. \text{ By Assumption 4, for any } t_i \text{ and } t'_i \text{, there exits a ball } \theta_j(t_i, t'_i) \text{ with radius } \phi \text{ such that } f(t_j|t_i) \neq f(t_j|t'_i) \text{ for all } t_j \in \theta_j(t_i, t'_i). \text{ Let } c_j(t_i, t'_i) \text{ be the center of } \theta_j(t_i, t'_i). \text{ Since } \beta\sqrt{d_i} < \phi, \text{ the hypercube containing } c_j(t_i, t'_i) \text{ is contained in } \theta_j(t_i, t'_i). \text{ Since the side length of each hypercube is } \beta, \text{ there is a ball of radius } \beta/3 \text{ within this hypercube such that } f(t_j, t_i) \neq f(t_j, t'_i) \text{ for all } t_j \text{ within the ball of radius } \beta/3. \text{ This defines a finite set of disjoint balls } B^{ij} \text{ satisfying the conditions of the lemma.}

Figure 2 illustrates. The large circle is the ball } \theta_j(t_i, t'_i) \text{ of radius } \phi \text{ given by Assumption 4. Point } c \text{ is the center of this ball. The grid size is } \beta, \text{ and so the maximum distance between two points in the same grid element is } \beta\sqrt{d_i}. \text{ This implies that the maximum distance, } r_m, \text{ between point } c \text{ and any other point in its grid element is less than } \beta\sqrt{d_i}, \text{ and hence less than } \phi. \text{ Therefore the grid element containing point } c \text{ is contained entirely within } \theta_j(t_i, t'_i), \text{ as is any ball contained entirely within this grid element. The dashed circle indicates one such ball. Since } f(t_j|t_i) \neq f(t_j|t'_i) \text{ for all } t_j \text{ in } \theta_j(t_i, t'_i), \text{ the same is true for all } t_j \text{ in the dashed circle. Since there are a finite number of grid elements, taking one such ball for every grid element gives a finite set of balls such that for each distinct } t_i \text{ and } t'_i \text{ there is at least one ball on which } f(t_j|t_i) \text{ and } f(t_j|t'_i) \text{ are not equal.}

**Proof of Theorem 2:** The proof will employ transfers based on a large scaling of the quadratic scoring rule. However, rather than working with } K_i \text{ as the (large) scaling applied to the transfers, yielding payoffs } u_i(a, t) = v_i(t, a) + K_i x_i(a_i, a_j), \text{ we will instead work with the equivalent formulation in which } \gamma_i = 1/K_i \text{ and payoffs are given by } \hat{u}_i(a, t) = \gamma_i v_i(t, a) + x_i(a_i, a_j). \text{ In this formulation, the agent’s utility function depends continuously on } \gamma_i. \text{ The game with } \gamma_i = 0 \text{ is one in which agent } i \text{ cares only about the transfer, and the game with } \gamma_i \text{ positive but small (which corresponds a large scaling of the transfers) can be thought of as a slightly perturbed version of the } \gamma_i = 0 \text{ game. The proof exploits the fact that if the conditions for application of Schauder’s}
fixed point theorem are satisfied when $\gamma_i = 0$ and best response correspondences are single-valued and suitably continuous, then they are also satisfied for small-but-positive values of $\gamma_i$.

Without loss of generality, assume $\delta \leq \delta^*$ as specified in Lemma 4. For each $i$ choose a player $j$ satisfying Lemma 4, and let $\tilde{B}^{ij} = \{\tilde{b}^{ij}_0, \tilde{b}^{ij}_1, ..., \tilde{b}^{ij}_M\}$, denote the partition of agent $j$’s announcement space described above. Let

$$p^{s_j(\cdot)}(\tilde{b}^{ij}_m|a_i) = \int_{\{t_j|s_j(t_j) \in \tilde{b}^{ij}_m\}} f(t_j|a_i) dt_j$$

be the probability of event $\tilde{b}^{ij}_m$ if agent $i$’s type is $a_i$, conditional on $j$’s announcement strategy. If agent $j$ plays strategy $s_j(\cdot)$, let the transfer to agent $i$ be $x^{s_j(\cdot)}_i(a_i, a_j)$, which is based on the quadratic scoring rule applied to the events in $\tilde{B}^{ij}$ according to:

$$x^{s_j(\cdot)}_i(a_i, a_j) = 2p^{s_j(\cdot)}(\tilde{b}^{ij}_m|a_i) - \sum_{m=1}^{M} p^{s_j(\cdot)}(\tilde{b}^{ij}_m|a_i)^2.$$ 

Note that for any $s_j(\cdot) \in C_j(\delta)$, by Lemma 4, transfers $x^{s_j(\cdot)}_i(a_i, a_j)$ represent a strictly proper scoring rule, and hence for each $i$, $s_j(\cdot) \in C_j(\delta)$, and each $t_i$ agent $i$’s expected transfer is maximized by announcing truthfully ($a_i = t_i$).

For any $s_j(\cdot)$, for each value of $t_i$ agent $i$ chooses $a_i$ to maximize:

$$\gamma_i \int_{T-i} v_i(t, s_{-i}(t_{-i}), a_i) f(t_{-i}|t_i) dt_{-i} + \int_{T_j} x^{s_j(\cdot)}_i(a_i, a_j) f(t_j|t_i) dt_j.$$

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20 If not, use $\delta^*$ in the following construction. Since it establishes that there is a $\delta^*$-truthful BNE, this BNE is also $\delta$-truthful.

21 We denote the player as $j$ rather than $j(i)$ for notational convenience.
Since $p^{s_j}_{ij}$ is continuous in $s_j (\cdot)$ (which implies that $x^{s_j}_{ic}(a_i,a_j)$ is continuous in $s_j (\cdot)$) and $v_i (t,a)$ is continuous in $t$ and $a$, (5) is continuous in $s_{-i} (\cdot)$, and in particular in $s_j (\cdot)$. Further, our assumptions ensure that (5) is twice continuously differentiable in $a_i$. Note that the transfers used are constructed for each $s_j (\cdot)$ in order to render truthful reporting a best response when $\gamma_i = 0$ (and thus nearly truthful reporting a best response when $\gamma_i$ is positive but sufficiently small).

Let $BA_i (t_i, s_{-i} (\cdot), \gamma_i)$ denote agent $i$’s best announcements (i.e., announcements that maximize (5)) given his type, the other agents’ strategies, and the weight $i$ places on his direct return. $BA_i (t_i, s_{-i} (\cdot), \gamma_i)$ may be multi-valued. Let $BR_i (s_{-i} (\cdot), \gamma_i)$ denote the best response correspondence for agent $i$. That is, $BR_i (s_{-i} (\cdot), \gamma_i)$ maps $s_{-i} (\cdot)$ and $\gamma_i$ to best-action correspondences $BA_i (t_i, s_{-i} (\cdot), \gamma_i)$. Let $BR (s (\cdot), \gamma) = (BR_1 (s_{-1} (\cdot), \gamma_1), ..., BR_N (s_{-N} (\cdot), \gamma_N))$ denote the best response correspondence for all agents, where $\gamma = (\gamma_1, ..., \gamma_N)$. The goal is to establish the existence of a $\delta$-truthful fixed point of this operator. Such a fixed point is a $\delta$-truthful BNE of the announcement game.

The remainder of the proof proceeds in three steps that adapt the steps in Meirowitz’s (2003, Proposition 1) existence result for general Bayesian games with infinite type and action spaces to the present case. Step 1 establishes that under the transfers specified above, when $\gamma_i$ is sufficiently small, each agent’s best response correspondence is single-valued and continuous, and thus that the best response operator is a continuous map of the space $C (\delta)$ into itself. Step 2 establishes that the range of the best response operator is uniformly bounded and equicontinuous, and thus that the best response operator is compact. In Step 3 we apply Schauder’s fixed-point theorem to establish the existence of a $\delta$-truthful BNE of this game.

**Step 1:** Note that for each $i$, $T_i$ is a compact, convex subset of a finite-dimensional Euclidean space with a non-empty interior, and that agent $i$’s objective function (5) is continuous in $t_i, a_i, s_{-i} (\cdot)$, and $\gamma_i$. Thus for any $t_i, \gamma_i$, and $s_{-i} (\cdot) \in C_{-i} (\delta)$, agent $i$’s best response exists and by the Theorem of the Maximum (Berge, 1997) $BA_i (t_i, s_{-i} (\cdot), \gamma_i)$ is an upper hemi-continuous function of $t_i$.

Next, we show that for $\gamma_i$ sufficiently small, $BA_i (t_i, s_{-i} (\cdot), \gamma_i)$ is single-valued for any $s_{-i} (\cdot) \in C_{-i} (\delta)$. Fix $s_{-i} (\cdot)$. Let $W^s_{ij} (a_i, t_i) = \int_{T_j} x^{s_j}_{ij} (a_i, a_j) f (t_j | t_i) d t_j$. Since these transfers are strictly proper, for any $t_i, a_i = t_i$ is the unique maximizer of $W^s_{ij} (a_i, t_i)$, and so $a_i = t_i$ solves the first-order necessary conditions: $D_{a_i} W^s_{ij} (t_i, t_i) = 0$. Letting $s^*_i (t_i) \equiv t_i$ and implicitly differentiating the first-order conditions with respect to $t_i$ and evaluating at $s^*_i (t_i) \equiv t_i$ yields the matrix equation:

$$
\begin{align*}
\left[ \frac{\partial^2 W^s_{ij}(t_i, t_i)}{\partial a_{ir} \partial a_{ic}} \right] \left[ \frac{\partial s^*_i}{\partial t_{ic}} \right] &= \left[ \frac{\partial^2 W^s_{ij}(t_i, t_i)}{\partial t_{ir} \partial t_{ic}} \right],
\end{align*}
$$

where $\left[ \frac{\partial^2 W^s_{ij}(t_i, t_i)}{\partial a_{ir} \partial a_{ic}} \right]$ denotes the $d_i \times d_i$ matrix with $\frac{\partial^2 W^s_{ij}(t_i, t_i)}{\partial a_{ir} \partial a_{ic}}$ in the $r^{th}$ row and $c^{th}$ column and $\left[ \frac{\partial^*_i}{\partial t_{ic}} \right]$ are similarly defined $d_i \times d_i$ matrices. Since $s^*_i (t_i) \equiv t_i$, this system has a unique solution $\left[ \frac{\partial^*_i}{\partial t_{ic}} \right] = I$ (where $I$ denotes the $d_i \times d_i$ identity matrix). Therefore $\det \left[ \frac{\partial^2 W^s_{ij}(t_i, t_i)}{\partial a_{ir} \partial a_{ic}} \right] \neq 0$. Applying the Implicit Function Theorem, for any $s_{-i} (\cdot) \in C_{-i} (\delta)$ and any $t_i$ there is an open ball of $\gamma_i$ around $\gamma_i = 0$ on which $BA_i (t_i, s_{-i} (\cdot), \gamma_i)$ is single-valued and a continuously differentiable function of $t_i$ and $\gamma_i$. (Differentiability is used in Step 2.)
Since $BA_i(t_i, s_{-i}(.), \gamma_i)$ is upper hemi-continuous and single-valued in $t_i$ for any $s_{-i}(.) \in C_{-i}(\delta)$ and $\gamma_i$, this implies that $BA_i(t_i, s_{-i}(.), \gamma_i)$ is a continuous function of $t_i$, and hence that $BR_i(s_{-i}(.), \gamma_i)$ maps to continuous functions of $t_i$ for $\gamma_i$ sufficiently small: $BR_i(s_{-i}(.), \gamma_i) : C_{-i}(\delta) \rightarrow C_i$. Further, since (5) depends continuously on $s_{-i}(.)$ and $\gamma_i$, the Theorem of the Maximum also establishes that $BR_i(s_{-i}(.), \gamma_i)$ is continuous in $s_{-i}(.)$ and $\gamma_i$ (since $BR_i(s_{-i}(.), \gamma_i)$ is upper hemi-continuous and single-valued).

Finally, we show that for $\gamma_i$ sufficiently small, best responses are close to truthful. Since (5) is continuous in $\gamma_i$, for $\gamma_i$ sufficiently small, for any $t_i$, $\|BA_i(t_i, s_{-i}(.), \gamma_i) - t_i\|_R < \delta$. Since we above established that $BR_i(s_{-i}(.), \gamma_i)$ is a continuous function of $t_i$, it follows that $BR_i(s_{-i}(.), \gamma_i) \in C_i(\delta)$ for $\gamma_i$ sufficiently small. For each $i$, let $\gamma_i^*$ be the largest $\gamma_i$ such that $BR_i(s_{-i}(.), \gamma_i) \in C_i(\delta) \cap C_i^1$ for all $s_{-i}(.) \in C_{-i}(\delta)$, where $C_i^1$ is the space of continuously differentiable functions from $T_i$ to $T_i$. Let $\gamma^* = \min_i \gamma_i^*$. Thus, when $\gamma_i < \gamma^*$, the best response operator is continuous and $BR(s(\cdot), \gamma) : C(\delta) \rightarrow C(\delta)$. For the remainder of the proof, we restrict ourselves to $\gamma < \gamma^*$.

Step 2: We next show that $BR(s(\cdot), \gamma)$ is a compact operator on $C(\delta)$. An operator is compact if it is continuous and maps bounded sets to relatively compact sets (Zeidler, 1985, p. 53). Step 1 establishes continuity. By the Arzela-Ascoli theorem, $BR(C(\delta), \gamma)$ is relatively compact if for any $D \subset C(\delta)$, $BR(D, \gamma) \equiv \{BR(\eta, \gamma) : \eta \in D\}$ is uniformly bounded and equicontinuous.\(^{22}\) Uniform boundedness requires that $\sup_{\eta \in BR(D, \gamma)} \left(\sup_{t \in T} ||\eta(t)||_{\rho}\right) < \infty$, where $||\cdot||_p$ denotes the product norm on $T$.\(^{23}\) Uniform boundedness is straightforward since for each $i$, $T_i$ is compact and $BR_i(s(\cdot), \gamma)$ is continuous in $t_i$. Thus $C(\delta)$ is uniformly bounded, and so any subset of $C(\delta)$ is uniformly bounded. To establish equicontinuity, we must show that for any $\phi_i > 0$ there exists $\psi_i(\phi_i) > 0$ such that

$$\sup_{s_i(.) \in BR_i(D, \gamma)} \left\|s_i\left(t_i^\prime\right) - s_i\left(t_i^\prime\prime\right)\right\|_R < \phi_i \text{ whenever } \left\|t_i^\prime - t_i^\prime\prime\right\|_R < \psi_i(\phi_i).$$

Since $BA_i(t_i, s_{-i}(.), \gamma_i)$ is continuously differentiable in $t_i$ and $\gamma_i$ for $\gamma_i < \gamma^*$ and $BA_i(t_i, s_{-i}(.), 0) \equiv t_i$, $\gamma_i$ can be chosen sufficiently small as to uniformly bound the $t_i$ derivatives of $s_i(\cdot) \in BR_i(C(\delta), \gamma)$, from which equicontinuity follows. Let $\gamma_i^{**}$ be such that equicontinuity holds for $\gamma_i < \gamma_i^{**}$, and $\gamma^{**} = \min_i \gamma_i^{**}$. For the remainder of the proof, consider only $\gamma_i = \gamma^{***} \leq \min\{\gamma^*, \gamma^{**}\}$. For such $\gamma^{***}$, $BR(s(\cdot), \gamma^{***})$ is a compact operator.

Step 3: Schauder’s Fixed Point Theorem states that a compact operator that maps a nonempty, closed, bounded, convex subset of a Banach space into itself has a fixed point.\(^{24}\) Clearly, $C(\delta)$ is nonempty, closed, bounded, and convex, and therefore Schauder’s theorem applies. The fixed point of the best-response mapping is a $\delta$-truthful BNE of this game, and thus a $\delta$-truthful BNE of the game where payoffs are given by $u_i(a, t) = v_i(t, a) + K_i x_i(a_i, a_j)$ for $K_i = 1/\gamma^{***}$. Transfers are balanced using the same type of permutation as was employed in the proof of Theorem 1.

\(^{22}\) See Meirowitz (2003) for a version of the theorem tailored to this environment, or Zeidler (1985, p. 772) for a more general version.

\(^{23}\) Since $T$ is compact, uniform boundedness implies that $\{\eta(\cdot) : \eta \in D\}$ is relatively compact in $T$.

\(^{24}\) See Zeidler (1985), Theorem 2A (Schauder’s Fixed Point Theorem (1930), p. 56).