Mechanism Design with Costly Information Acquisition Part IV of Group Decision Making

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Theory Reading Group

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Mechanism Design with Costly Information A

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- Introduction
- @ Gerardi & Yariv, GEB 2008, "Information Acquisition in Committees"
- Gershkov & Szentes, JET 2009, "Optimal Voting Schemes with Costly Information Acquisition"
- Discussion (online environment)

Information is not always free. Sometimes it is costly to acquire information necessary to good decision-making.

True of both individual and group decisions.

Optimal group decision-making procedures need to account for the fact that agents can choose whether or not they are informed.

• Groups of homogeneous individuals (common goals) with costly information acquisition

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- Seeking optimal decision-making procedure (normative exercise)
- Free-rider problem (problem with large groups)
- Tradeoff between informed decision making (aggregation) and providing incentives to invest in information (requires making individual agents' information more important).

- **GY**: SP chooses committee size and decision rule, agents simultaneously choose investment and report.
 - Main Result: The ex-ante optimal mechanism may be ex-post inefficient.
- **GS**: Restrict attention to ex-post efficient mechanisms.
 - **Main Result**: Requires sequentiality. After each individual agent reports his type, SP decides whether or not to continue collecting information. Precision threshold for stopping decreases with number of signals already acquired.

Gerardi & Yariv 2008 Model

Typical jury setup, $N \ge 2$

Two states $\Omega = \{I, G\}$, they occur with probability P(I), P(G)Decisions $\{A, C\}$

 $u(d, \omega)$ given by u(C, I) = -q u(A, G) = -(1-q) $u(d, \omega) = 0$ otherwise

Each agent can purchase a single signal $s \in \{i, g\}$ of accuracy $p > \frac{1}{2}$ by paying a cost c > 0

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SP chooses extended mechanism

- committee size n ≤ N (assume chooses smallest given level of feasible expected payoffs)
- ② symmetric mapping γ : {0, 1, ..., n} → [0, 1], $\gamma(k) \equiv \Pr(d = C \mid \#_g = k)$
- Agents observe mechanism, and j = 1, ..., n simultaneously decide whether to purchase signal (restrict attn to pure strategies and collection)
- Each agent sends a message in {i, g} to SP, who uses mechanism to select outcome

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Gerardi & Yariv 2008

Incentive Compatibility

$$\sum_{k=0}^{n-1} {\binom{n-1}{k}} f(k+1;n)(\gamma(k+1) - \gamma(k)) \ge c \qquad (IC_i)$$

$$\sum_{k=0}^{\infty} {\binom{n-1}{k}} f(k;n)(\gamma(k) - \gamma(k+1)) \ge c \qquad (IC_g)$$

where

$$f(k;n) = -qP(I)(1-p)^{k}p^{n-k} + (1-q)P(G)p^{k}(1-p)^{n-k}$$

proxies for difference in payoff between C and A when k of n signals are g

Constraints weight difference in payoff by conviction probabilities given profiles of signals

Define k_n smallest number of signals g such that SP weakly prefers C

Gerardi & Yariv 2008

The Optimization Problem

SP's problem:

$$\max_{\gamma:\{0,...,n\}\to[0,1]} Eu = -(1-q)P(G) + \sum_{k=0}^{n} {n \choose k} f(k;n)\gamma(k) \quad (P_n)$$

s.t. $(IC_i), \ (IC_g)$

 $\bar{\gamma}_n$ is sol'n to P_n if it exists, V(n) is explutive of optimal device

 n^* is optimal size, $V(n^*) \geq V(n) \,\, orall \,\, n \leq N$

Consider optimal devices $(n^*, \bar{\gamma}_{n^*})$

Note: SP does not take cost of information into account

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 γ_n^B is sol'n to maximization problem without constraints

$$\begin{aligned} \gamma_n^{\mathcal{B}}(k) &= 0 \text{ if } k < k_n \\ &= 1 \text{ if } k \ge k_n \end{aligned}$$

Define $\hat{V}(n)$ the expected utility of first best

Proposition 1: For every $n \ge 1$, $V(n) = \hat{V}(n)$ iff $\hat{V}(n) - \hat{V}(n-1) \ge c$ (Because if n-1 have been truthful, last player can guarantee $\hat{V}(n-1)$ by simply saying *i* or *g*) Assume there is at least one committee size for which Bayesian device admissable (n^B denotes greatest: $\hat{V}(n^B) - \hat{V}(n^B - 1) \ge c$, but $\forall n > n^B$, $\hat{V}(n) - \hat{V}(n-1) < c$)

SP can induce more than n^B to acquire info only if mechanism aggregates available signals suboptimally

But more information is available in larger committees.....key tradeoff

Proposition 2: Fix P(I), q, p and assume regular environment (tech assm). Let $n^*(c) \leq N$ denote the optimal committee size when cost is c. $\exists \ \bar{c} > 0 \ s.t. \ \forall \ c < \bar{c}$, whenever $\hat{V}(N) - \hat{V}(N-1) < c$, then $V(n^*(c)) < \hat{V}(n^*(c))$ (i.e. optimal device is not Bayesian)

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Gerardi & Yariv 2008

Intuition of Main Result

- Consider Bayesian device with $n^B + 1$ agents
 - Difference in payoffs when exactly k_{n^B} others report g but agent sees signal i
 - By definition, $\hat{V}(n^B+1) \hat{V}(n^B) < c$
 - Agent has incentive to save cost c by just reporting g (or i)
- Consider device with k_{n^B}
 - \bullet Difference in payoffs comes when exactly k_{n^B-1} others report g and agent sees i
 - $\tilde{V}(n^B) \hat{V}(n^B 1) \ge c$
- Consider combination
 - A when $k < k_{n^B}$, C when $k \ge k_{n^B+1}$, C with prob lpha when $k = k_{n^B}$
- Distorted device can have strictly higher payoff than Bayesian device, and induces $n^B + 1$ agents to acquire signals (Preview of GS results for ex ante optimal sequential mechanisms)

- Optimal device ex ante may be ex post inefficient
- Optimal distortions to ex post efficient rule depend on accuracy of signals (can be described for extreme p)
- Omparative statics on c, p
 - expected utility monotonic in c, p for given n
 - if restrict to ex-post efficient mechanisms, optimal committee size non-monotonic in *p* (possible to prefer many uninformative signals to few accurate signals)

Key differences:

- Committment to ex-post inefficient decisions problematic; look for optimal ex post efficient mechanism
- SP doesn't have to commit to a committee size at outset
- SP cares about information costs incurred by agents

Main Result: sequentially ask agents to collect info and vote; use decreasing decision threshold

 $K \ge 1$ individuals

Two states: A and B, equiprobable (robust to some asymmetry)

Decisions: α and β , where

$$\begin{array}{ll} u(\alpha & \mid A) = u(\beta \mid B) = 1 \\ u(\alpha & \mid B) = u(\beta \mid A) = 0 \end{array}$$

(where again, robust to asymmetry)

Agents can purchase a signal in $\{a, b\}$ of accuracy p at cost c. This takes no time (i.e. no discounting).

SP maximizes $K_0 Eu - c\bar{L}$ (\bar{L} = expected number of agents who collect info)

Notation: s is a finite sequence of signals; d is the difference in number of a and b signals in s

Note: Ex post efficient decision is to follow the majority of signal reports

Proposition 1: There exists a weakly decreasing (in single steps) function $g: \mathbb{N} \to \mathbb{N}$ s.t. if, after asking I agents, the reported signal sequence is such that |d| > g(I), the SP makes the majority decision. Otherwise, SP asks an additional agent.

• If $K = \infty$ then there exists $k \in \mathbb{N}$ such that $g \equiv k$. In addition, $k \to \infty$ as $K_0 \to \infty$.

② If
$$K < \infty$$
 then for all $l \in \mathbb{N}$, $g(l+1) = g(l)$ or $g(l+1) = g(l) - 1$, and $g(K-1) = 1$.

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Minor results towards main result

- Restrict attn to sequential mechs in which agents act only once, which are ex post efficient and symmetric wrt A and B, which uniformly order agents for all K < ∞</p>
- Can explicitly characterize IC constraint (both for acquiring signal and truthfully reporting it)
- Sestrict attention to Markovian mechanisms
 - A state V(I, d) ≡ {s | I(s) = I, d(s) = d}; Markovian mechanisms specify a (possibly random) decision D ∈ {M, (m), C} for each state V(I, d)
 - O The optimal mechanism generically involves randomization

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Looking for optimal Markovian mechanism

$$\begin{split} \max_{G} K_{0} & \sum_{V \in V(K)} [\rho(V, M : G)P(|d(V)|) + \rho(V, m : G)Q |d(V)|] \\ & -c\bar{L}(G) \\ \text{s.t.} & \frac{p-q}{4pq} & \sum_{V \in V(K)} \rho(V, M : G)P(|d(V)|)[|d(V)| - (p-q)I(V)] \\ & + \frac{p-q}{4pq} & \sum_{V \in V(K)} \rho(V, m : G)Q(|d(V)|)[-|d(V)| - (p-q)I(V)] \ge c\bar{L} \end{split}$$

Continuation Mechanisms (c.m.)

– $G(\mathit{I}, \mathit{d})$ a Markovian mechanism that follows reaching $V(\mathit{I}, \mathit{d})$ – note interaction through IC constraint

– Efficiency of mechanisms: $e(G(I, d)) = \frac{\Delta W(G(I,d))}{|\Delta I C(G(I,d))|}$ a measure of the tradeoff between effects on objective function and IC constraint

- 4. Use of a c.m. must increase the posterior in expectation
- 5. Monotonicity: for a given posterior, a longer sequence implies less efficient to continue asking more
- 6. If a c.m. used with positive probability in some state, can't stop in another state if a more efficient c.m. is available
- 7. Don't acquire more info if it can't possibly change the decision
- 8. Either no randomization or randomization only in $V(\hat{l}, |\hat{d}|)$

Gershkov & Szentes 2009 Main Result

Theorem 1: Suppose that the first-best mechanism does not satisfy the incentive compatibility constraint. Let G^* be an ex ante optimal mechanism among the ex post efficient ones. Then, there exists a decreasing step function $f : \mathbb{N} \to \mathbb{N}$, and $N \in \mathbb{N}$, such that

for all
$$l \in \mathbb{N}$$
, $f(l+1) = f(l)$ or $f(l+1) = f(l) - 1$, and $f(N) = 1$
et

$$T = \{V : f(I(V)) = |d(V)|, f(I(V) - 1) = f(I(V)) + 1\}.$$

 G^{\star} is defined by the following three conditions:

1 If $V \notin T$ and $f(I(V)) \le |d(V)|$ then $p(M : V, G^*) = 1$ **2** If $V \notin T$ and f(I(V)) > |d(V)| then $p(C : V, G^*) = 1$ **3** If $V \in T$ then $p(M : V, G^*) \ge 0$, $p(C : V, G^*) > 0$

Furthermore, generically, there exists an optimal ex post efficient mechanism for which there are only two states, V_1 , $V_2 \in T$, such that $p(M : V_i, G^*) > 0$ for i = 1, 2. In addition, if $K < \infty$, then N = K.

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Optimal mechanism (if first best not incentive compatible) is to:

- Ask agents sequentially to acquire and report signals, without letting them know their position or any previous reports
- Stop acquiring and make efficient decision when posterior exceeds a cutoff, which is decreasing (weakly) in the number of reports
 - The cutoff jumps down by at most 1 for an additional report
- There is a bound $N \leq K$ s.t. SP never asks more than N+1 agents
- Generically, the mech will involve randomization
 - But only in relatively few states that satisfy certain criteria

Note: theorem holds even when $K = \infty$ (solely through the IC constraint)

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Gershkov & Szentes 2009

Sequential vs. Simultaneous

Sequential mechanism beats out simultaneous mechanism through:

Cost efficiency

- Example: $K = K_0 = 9$, p = 2/3, c = 0.04
- Simultaneous asks 5. If s = (a, a, a), SP incurs cost of two signals which won't change the decision
- Sequential stops asking when cost of information exceeds expected benefit (*s* = (*a*, *a*, *a*, *b*, *a*))
- What if introduce discounting?
- 2 Efficient incentive provision
 - Agents acquire information only if sufficiently likely that they are pivotal
 - Provide opportunities to be pivotal by sometimes acting on imprecise posteriors
 - Decreasing threshold means that acting on imprecise posteriors occurs more often after long sequences than short
 - Exploit difference in actual and conditional probability of long sequences to provide incentive at low cost

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Ex ante optimal mechanisms

- The ex ante optimal mechanism sometimes involves ex post inefficiency (as in GY)
- Ex ante optimal mechanism is also ex post efficient provided p is small enough (for a given K)

Issues from Ronny's introduction

- Information aggregation here not just concerned with aggregating, but with generating information for better decision-making
- What's bad about large groups? free rider problem
- Congestion here time constraints not an issue (info collected instantaneously); if this were not true, risk of decisions taking a long time (issue of discounting), but also ignorance of position in sequence becomes questionable (back to simultaneous?)
- Attention constraints not addressed, but worth noting computation intensity
- Sequentiality crucial in GS, but not for usual reasons