

Linear Contracting Models with Multiple Efforts

- Multiple tasks $e_1 \dots e_n$.

– Output:

$$\begin{aligned} x_i &= e_i + \theta_i \\ \theta &\sim N(0, \Sigma) \end{aligned}$$

Σ allows for correlation between tasks

– Cost:

$$c(e) = c(e_1, \dots, e_n)$$

– Principal has some benefit function $B(e)$.

– Contracts:

$$s(x) = \alpha x + \beta$$

– Note: CE in normal exp. model is

$$\alpha e + \beta - \frac{1}{2} r \alpha' \Sigma \alpha - c(e)$$

β shifts surplus \Rightarrow linear Pareto frontier

– TCE:

$$B(e) - c(e) - \frac{1}{2} r \alpha' \Sigma \alpha$$

$$\max_{\alpha} \quad B(e) - c(e) - \frac{1}{2} r \alpha' \Sigma \alpha$$

$$\text{s.t. } \alpha = \nabla_e c(e) \tag{IC}$$

$$\text{i.e., } \frac{\partial c}{\partial e_i} = \alpha_i.$$

Observe:

$$D\alpha(e) = D^2c(e) = \begin{bmatrix} \frac{\partial^2}{\partial e_1^2}c(e) & \cdots & \frac{\partial^2}{\partial e_1 \partial e_n}c(e) \\ \vdots & \ddots & \vdots \\ \frac{\partial^2}{\partial e_n \partial e_1}c(e) & \cdots & \frac{\partial^2}{\partial e_n^2}c(e) \end{bmatrix}$$

FOC:

$$\begin{aligned} \nabla B(e) &= \nabla c(e) + r D^2c(e) \Sigma \nabla c(e) \\ &= (I + r D^2c(e) \Sigma) \nabla c(e) \\ \alpha^* &= (I + r D^2c(e) \Sigma)^{-1} \nabla B(e) \end{aligned}$$

- Case 1:

- Outputs stochastically independent, efforts independent in cost function.

$$\begin{aligned} \Sigma_{ij} &= 0, \text{ for } i \neq j \\ \frac{\partial^2 c}{\partial e_i \partial e_j} &= 0, \text{ for } i \neq j \end{aligned}$$

- This yields

$$\alpha_i = \frac{\partial B / \partial e_i}{1 + r \frac{\partial^2 c}{(\partial e_i)^2} \sigma_i^2}$$

- Case 2:

- Unmonitored tasks: two tasks ($c(x)$ convex)

$$\sigma_{12} = 0; B_{12} = 0; c_{12} = 0; \sigma_2^2 \rightarrow \infty$$

$$\alpha_1 = \frac{B_1 - \frac{c_{12}}{c_{22}}B_2}{1 + r\sigma_1^2(c_{11} - \frac{c_{12}^2}{c_{22}})}$$

$$(c_{11} - \frac{c_{12}^2}{c_{22}}) > 0 \quad (\text{convexity})$$

- $\alpha_2 = 0 \rightarrow$ no incentive for task 2 (too expensive)
- If $c_{12} > 0$, $B_2 > 0$, then $\alpha_1 \downarrow$
- Incentives to activity 1 are also decreasing
- Suppose:

$$\begin{aligned} c(e) &= \hat{c}(e_1 + e_2) \\ B(e_1, 0) &= 0 \quad \forall e_1 \end{aligned}$$

Then, as $\sigma_2^2 \rightarrow \infty$, $\alpha_2 \rightarrow 0$.

- If $\alpha_1 > 0$, then $e_2 = 0$, leading to no benefit to the principal ($B = 0$).
- Hence if $\alpha = (0, 0)$ induces any effort it is better.
- Justification for low-powered incentives!

- Holmstrom and Milgrom, AER 1994: “Firm as an Incentive System”

- Suppose task 2 is asset maintenance (machines, firm reputation)
- Solution: $\alpha_2 = 1$ using ownership
- Two systems:
 - * high-powered incentives, ownership
 - * “employee”–firm owns assets, fixed wages
- Under what conditions are α_1 , α_2 complements in the principal’s objective function?

$$TCE = TCE(\alpha_1, \alpha_2, \sigma_1^2, \sigma_2^2, \sigma_{12}, r)$$

- Assumptions on the cost function: assume quadratic for simplicity.
 - * Efforts substitutes in the cost function: $c_{12}(e_1, e_2) \geq 0$.
 Implies $\frac{\partial^2}{\partial \alpha_1 \partial \alpha_2} TCE \geq 0$ if c is quadratic.
 - * Efforts complements in the cost function: $c_{12}(e_1, e_2) \leq 0$.
 Implies $\frac{\partial^2}{\partial \alpha_1 \partial \alpha_2} TCE \leq 0$ if c is quadratic.
- Examples of questions to ask in this framework:
 - What happens when σ_1^2 changes?
 - How does the choice of $\alpha_1, \alpha_2, \sigma_1^2$ vary with the cost of changing technology?
- Intuition: when efforts are substitutes in the cost function, incentive instruments are complements to the principal.

Comparative Statics Detour

Comparative statics tools: Need for firm as a system and for mechanism design, as well as Hart-Moore 1990.

$$x^*(\theta, S) = \arg \max_{x \in S} f(x, \theta)$$

- When is $x^*(\theta, S)$ increasing in θ ?

Question: Does a solution exist? Are there multiple optima?

$x^*(\theta, S)$ could be empty or a set

- What is an order?

- partial order: $x, y \in X$, could have $x \not\geq y$ and $y \not\geq x$
- Componentwise Order on \mathbb{R}^n : $a \geq b$ iff $a_i \geq b_i; \forall i = 1, \dots, n$

* Notation: Meet and Join

$$a \vee b = (\max(a_1, b_1), \dots, \max(a_n, b_n))$$

$$a \wedge b = (\min(a_1, b_1), \dots, \min(a_n, b_n))$$

- versus complete order
 - * e.g., lexicographic order, usual order on real line
- Orders Over Sets
 - * Inclusion order: $A \geq_I B$ iff $B \subseteq A$

- – * Strong Set Order, Case 1: $A, B \subset \mathfrak{R}$
 - $A \geq_{SSO} B \Leftrightarrow$
 $a \in A \text{ and } b \in B \Rightarrow \max(a, b) \in A, \min(a, b) \in B$
 - Equivalent definition: $x \in B \setminus A, y \in A \cap B, z \in A \setminus B \Rightarrow$
 $x \leq y \leq z$
- * Case 2: $A, B \subset \mathfrak{R}^n$
 - $A \geq_{SSO} B \Leftrightarrow$
 $a \in A \text{ and } b \in B \Rightarrow a \vee b \in A, a \wedge b \in B$
- * Fact: $x^*(\theta) \uparrow \theta$ in SSO

$$\Rightarrow \underbrace{x^H(\theta)}_{\text{highest member of } x^*(\theta)}, \underbrace{x^L(\theta)}_{\text{lowest member of } x^*(\theta)} \uparrow \theta$$

- General Definitions:

- Lattice: set X , partial order \geq
- “meet” \wedge , “join” \vee

$$x \vee y = \inf \{z : z \geq x, z \geq y\}$$

$$x \wedge y = \sup \{z : z \leq x, z \leq y\}$$
- Sublattice: $S \subseteq X$ such that $x \in S, y \in S$
 $\Rightarrow x \vee y \in S \text{ and } x \wedge y \in S$
- Supermodularity: $f : X \Rightarrow \mathbb{R}$ is SPM, if $\forall x, y \in X$,

$$f(x \vee y) + f(x \wedge y) \geq f(x) + f(y)$$

$$\Leftrightarrow f(x \vee y) - f(x \wedge y) \geq [f(x) - f(x \wedge y)] + [f(y) - f(x \wedge y)]$$

- Suppose $x = (z_1^H, z_2^L), y = (z_1^L, z_2^H)$

$$x \vee y = (z_1^H, z_2^H), \quad x \wedge y = (z_1^L, z_2^L)$$

Then,

$$f(x \vee y) - f(x) \geq f(y) - f(x \wedge y)$$

$$\Leftrightarrow$$

$$f(z_1^H, z_2^H) - f(z_1^H, z_2^L) \geq f(z_1^L, z_2^H) - f(z_1^L, z_2^L)$$

- Restated question: When is $x^*(\theta, S)$ increasing in θ in strong set order?

– Question: What if $x^*(\theta) = \emptyset$ for some θ ?

Fact: $\emptyset \geq_{SSO} A \geq_{SSO} \emptyset \forall A$.

- Theorem (Topkis, 1978): If $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is differentiable, it is SPM if $\forall i \neq j, \forall x$

$$\frac{\partial^2}{\partial x_i \partial x_j} f(x) \geq 0.$$

Proof: Apply definition of derivative.

- Interpretation:

- Increasing first component increases marginal returns to other components
- “Supermodular” \Leftrightarrow for product set, every pair of choices complements

- Theorem (Topkis, 1978): If $f(x, \theta)$ is SPM and S is a sublattice, then $x^*(\theta, S) = \arg \max_{x \in S} f(x, \theta)$ is nondecreasing in θ in the strong set order.

SPM is not necessary. We will discuss necessity later.

- Games:

Player payoffs: $f^1(x_1, x_2, \theta_1); f^2(x_1, x_2, \theta_2)$

Suppose for each i , f^i SPM in (x_1, x_2)

and, f^i is supermodular in components of x_i if x_i is a vector.

- Theorem (Topkis, 1979): If each f^i is SPM in (x_i, x_j, θ) , then eqm $(x_1^*, x_2^*) \uparrow \theta$.

- Applying the theorem:

– use θ to parametrize diff't games

$$\theta_L : f^i(x_1, x_2, \theta_L) = f(x_i, x_j) - c^i(x_i) - c^j(x_i)$$

$$\theta_H : f^i(x_1, x_2, \theta_H) = f(x_i, x_j) - c^i(x_i) \quad (\text{don't experience cost on other player})$$

– Social θ_L versus Nash Equilibrium θ_H

if $c^i, c^j \uparrow x$ then game has “too much” x

– See Ilya Segal, “Contracting with Externalities.”

Applying Comparative Statics to Multitask Problems

- Example: the two-effort linear contracting model.

The risk premium is:

$$-\frac{1}{2}r\alpha_1^2\sigma_1^2 - \frac{1}{2}r\alpha_2^2\sigma_2^2 - r\alpha_1\alpha_2\sigma_{12}$$

Verify that when c quadratic, $c_{12} \geq 0$, we have:

$$\frac{\partial^2}{\partial\alpha_1\partial\alpha_2} TCE(\alpha_1, \alpha_2, \sigma_i^2) \geq 0$$

Verify: $\frac{\partial^2}{\partial\alpha_1\partial\sigma_1^2} TCE \leq 0$, $\frac{\partial^2}{\partial\alpha_2\partial\sigma_1^2} TCE = 0$

- Comp Statics: I

- If $\sigma_1^2 \downarrow$ $\alpha_1 \uparrow$ $\alpha_2 \uparrow$
- If technology is a choice, then reinforcing choices: $\sigma_1^2 \downarrow$ $\alpha_1 \uparrow$ $\alpha_2 \uparrow$
- Suppose we observe different firms, and that (σ_1^2, σ_2^2) vary independently in population. Each firm t gets a draw of (σ_1^2, σ_2^2) .
- Since $(\alpha_1, \alpha_2) \downarrow (\sigma_1^2, \sigma_2^2)$, there will be a positive correlation between α_1 and α_2 .
- We expect to see some firms with low (α_1, α_2) and some with high $(\alpha_1, \alpha_2) \Rightarrow$ in general expect to see clusters.
 - * Some firms, however, will appear “mis-matched” due to high σ_1^2 and low σ_2^2 .

- Application: Franchising

- Royalty rate: proportion of profits paid to company
- Asset maintainance: mandatory advertising

- Monitoring technology: inspection for quality
 - Empirical prediction: company owned on interstate, franchises in local stores
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- Note: interpretations of α_1, α_2 can be quite general (i.e. government-owned or privately owned, etc.) See Hart, Schliefer, Vischny on prisons.
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- Common Agency (Holmstrom and Milgrom unpublished paper; also see Dixit)
 - Two principals
 - Two efforts
 - Case 1: Principal i cares about x_i , cannot observe x_j .
 - * “Overworked student”
 - * Exert externalities on other principal, over-incentivize agent
 - Case 2: Principal i cares about x_i , can observe both outputs.
 - * Low-powered incentives for politicians (Dixit)
 - * Incentive schemes cancel each other out