Organizations and Agents: Multi-tasks

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Lecture 11

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Multiple Tasks

Multiple Tasks I

So far, we modeled production wherein

- Agent performed only one task;
- There was only one output q.

In real world,

- employees at work perform multi-tasks
- produce several outputs

For example,

- Workers
 - Produce output (using firm's assets)
 - Maintain assets

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Multiple Tasks II

- Managers/CEO
 - Supervise existing workers/employees
 - Train existing workers/employees
 - Hire new workers/employees
- Salespersons
 - Promote sale with existing customers
 - Make new customers
 - Launch sale of new products
- Teachers
 - Teach
 - Research
 - Serve on administrative committees

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Multiple Tasks III

The output is also multi-dimensional

- Workers output
 - Quantity/units of output
 - Residual value of assets
- Managers/CEO
 - Current profits
 - Value of stocks/shares of company
- Teachers
 - Teaching quality and quantity
 - Research output

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Model I

Holmstrom and Milgrom (1991, J Law Eco and Organizations) A simple version:

- Two tasks; *i* = 1, 2
- Two signals/outputs: $q_i(e_i, \epsilon_i) = e_i + \epsilon_i$, i = 1, 2. Specifically, $q_i(e_i, \epsilon_i) = e_i + \epsilon_i$, where

$$\begin{array}{rcl} q_1(e_1,\epsilon_1) &=& e_1+\epsilon_1 \\ q_2(e_2,\epsilon_2) &=& e_2+\epsilon_2, \end{array}$$

 $\epsilon = (\epsilon_1, \epsilon_2) \sim \textit{N}(0, \Sigma),$ where Σ

• $\epsilon \sim N(0, \Sigma)$, where Σ is variance-covariance matrix;

$$\Sigma = \left(egin{array}{cc} \sigma_1^2 & R \ R & \sigma_2^2 \end{array}
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Model II

Payoffs:

- Contract: $w(q_1, q_2) = t + \sum_{i=1}^2 s_i x_i = t + \mathbf{s}^T \mathbf{q}$, where $s_i \ge 0$
- Principal is risk-neutral with expected payoff $V = V(\mathbf{q}, w) = V(q_1, q_2, w)$, i.e., $V = V(\mathbf{e}, w) = V(e_1, e_2, w)$
- Agent is risk-avers: $u(w, \mathbf{e}) = -e^{-r(w-\psi(\mathbf{e}))}, r > 0$, where

•
$$r = -\frac{u''}{u'} > 0$$
, i.e., CARA, and

• Principal's payoff: $V(q_1, q_2, w) = E(q_1 + q_2 - w) = e_1 + e_2 - E(w)$

•
$$\psi(e) = \frac{1}{2}c_1e_1^2 + \frac{1}{2}c_2e_2^2 + \delta e_1e_2$$

•
$$\psi_1(.) = \frac{\partial \psi(\mathbf{e}_1, \mathbf{e}_2)}{\partial \mathbf{e}_1} = \mathbf{c}_1 \mathbf{e}_1 + \delta \mathbf{e}_2$$
 and $\psi_2(.) = \frac{\partial \psi(\mathbf{e}_1, \mathbf{e}_2)}{\partial \mathbf{e}_2} = \mathbf{c}_2 \mathbf{e}_2 + \delta \mathbf{e}_1$. So

Model III

- $\left\{ \begin{array}{ll} \delta = 0 & \text{tasks are independent;} \\ \delta > 0 & \text{tasks are technological substitutes;} \\ \delta < 0 & \text{tasks are technological complements.} \end{array} \right.$

imperfect substitutes if $0 < \delta < \sqrt{c_1 c_2}$

• Contract: $w(q_1, q_2) = t + s_1q_1 + s_2q_2$, where $s_i \ge 0$. Note

$$E(w(q_1, q_2)) = E(t + s_1(e_1 + \epsilon_1) + s_2(e_2 + \epsilon_2))$$

= t + s_1e_1 + s_2e_2.

- $Var(t + s_1(e_1 + \epsilon_1) + s_2(e_2 + \epsilon_2)) = s_1^2 \sigma_1^2 + s_2^2 \sigma_2^2 + 2Rs_1 s_2$
- $\bar{w} = \text{Certainty equivalent of the reservation wage (the outside option for$ the agent)

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First Best

The first best is solution to

$$\max_{e_i} E(\sum q_i - w)$$

s.t. $-e^{-r[w - \psi(e_1, e_2)]} = -e^{-r\bar{w}}$, i.e., $w - \psi(e_1, e_2) = \bar{w}$, i.e.,
 $w = \bar{w} + \psi(e_1, e_2)$.

Therefore, the first best is solution to

$$\max_{e_1,e_2} E(e_1 + \epsilon_1 + e_2 + \epsilon_2 - \bar{w} - \psi(e_1,e_2)), i.e.,$$
$$\max_{e_1,e_2} \{e_1 + e_2 - [\frac{1}{2}c_1e_1^2 + \frac{1}{2}c_2e_2^2 + \delta e_1e_2]\}$$

Therefore, the first best efforts, e_1^* and e_2^* , solve the following foc

$$\psi_1(e) = c_1 e_1 + \delta e_2 = 1$$
 (0.1)

$$\psi_2(e) = c_2 e_2 + \delta e_1 = 1.$$
 (0.2)

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Multiple Tasks

Second Best I

e is not contractible but q is. As before, the agent solves

 $\max_{e_1,e_2} \{ \hat{w}(e_1,e_2) \},$

where



Second Best II

The foc w.r.t. e_1 and e_2 are

$$\mathbf{s}_1 = \mathbf{c}_1 \mathbf{e}_1 + \delta \mathbf{e}_2 \tag{0.3}$$

$$s_2 = c_2 e_2 + \delta e_1 \tag{0.4}$$

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That is,

$$\mathbf{s}(\mathbf{e}) = \nabla \psi(\mathbf{e}).$$

IR is given by

$$\begin{split} u(\hat{w}(e_1, e_2)) &\geq u(\bar{w}), i.e., \quad \hat{w}(e_1, e_2) \geq \bar{w}, i.e., \\ t + s_1 e_1 + s_2 e_2 - [\frac{1}{2}c_1 e_1^2 + \frac{1}{2}c_2 e_2^2 + \delta e_1 e_2] - \frac{r}{2}[s_1^2 \sigma_1^2 + s_2^2 \sigma_2^2 + 2Rs_1 s_2] \geq \bar{w} \quad (0.5) \\ \text{Fhe principal solves } \max_{e_1, e_2, t, s_1, s_2} E[q_1 + q_2 - w(q_1, q_2)], \text{ i.e.,} \\ \max_{e_1, e_2, t, s_1, s_2} E[q_1 + q_2 - (t + s_1 q_1 + s_2 q_2)], i.e., \end{split}$$

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Second Best III

 $\max_{e_1,e_2,t,s_1,s_2} E[e_1 + (1 - s_1)\epsilon_1 + e_2 + (1 - s_2)\epsilon_2 - (t + s_1e_1 + s_2e_2)]$

s.t. (0.3) - (0.5) hold. Clearly, (0.5) will bind. Therefore, the Principal's problem can be written as

$$\max_{e_1,e_2,s_1,s_2} \{e_1 + e_2 - [\frac{1}{2}c_1e_1^2 + \frac{1}{2}c_2e_2^2 + \delta e_1e_2] - \frac{r}{2}[s_1^2\sigma_1^2 + s_2^2\sigma_2^2 + 2Rs_1s_2]\},$$

s.t. (0.3) and (0.4) hold.

Note that the Principal programme can be written as

$$\max_{\mathbf{e}} \{ V(\mathbf{e}) - \psi(\mathbf{e}) - \frac{r}{2} \mathbf{s}^T \Sigma \mathbf{s} \}$$

s.t. $\mathbf{e} = \arg \max{\{\mathbf{s}^T \mu(\mathbf{e}) - \psi(\mathbf{e})\}}$

where $\mathbf{s}^T = (S_1, S_2)$. Ram Singh (DSE) Multiple Tasks 11/14

Second Best IV

Special Case 1: R=0

From (0.3) and (0.4) we get

$$e_{1} = \frac{s_{1}c_{2} - \delta s_{2}}{c_{1}c_{2} - \delta^{2}}$$
(0.6)
$$e_{2} = \frac{s_{2}c_{1} - \delta s_{1}}{c_{1}c_{2} - \delta^{2}}$$
(0.7)

The FOC w.r.t to s_1 is

$$s_{1} = \frac{c_{2} - \delta + \delta s_{2}}{c_{2} + r\sigma_{1}^{2}[c_{1}c_{2} - \delta^{2}]}$$
(0.8)

By symmetry FOC w.r.t. s_2 gives

$$s_{2} = \frac{c_{1} - \delta + \delta s_{1}}{c_{1} + r\sigma_{2}^{2}[c_{1}c_{2} - \delta^{2}]}, i.e., \qquad (0.9)$$

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Second Best V

From (0.8) and (0.9), we can see that $\frac{\partial s_i}{\partial \sigma_i} < 0$ and $\frac{\partial s_i}{\partial \sigma_j} < 0$. Further, in view of (0.8)

$$s_2^{SB} = \frac{1 + r\sigma_1^2(c_1 - \delta)}{(1 + r\sigma_1^2 c_1)(1 + r\sigma_2^2 c_2) - \delta^2 \sigma_1^2 \sigma_2^2 r^2}$$
(0.10)

Similarly,

$$s_1^{SB} = \frac{1 + r\sigma_2^2(c_2 - \delta)}{(1 + r\sigma_1^2 c_1)(1 + r\sigma_2^2 c_2) - \delta^2 \sigma_1^2 \sigma_2^2 r^2}$$
(0.11)

From (0.10) and (0.11), it can be checked that

$$rac{ds_i}{d\sigma_i} < 0 \; and \; rac{ds_i}{d\sigma_j} < 0.$$

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Second Best VI

That is, if it is hard to measure a task,

- low incentive is provided for that talk
- low incentive is provided even for the measurable task

Moreover, if production is very noisy, $\sigma_2^2 \Rightarrow \infty$ implies

$$\begin{aligned} s_2 &\Rightarrow & 0 \\ s_1 &\Rightarrow & \frac{r(c_2 - \delta)}{(1 + r\sigma_1^2 c_1)rc_2 - \delta^2 \sigma_1^2 r^2} \end{aligned}$$

That is,

• if it is impossible to measure a task, no incentive is provided for that talk

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