Justifying the First-Order Approach in Agency Frameworks with the Agent's Possibly Non-Concave Value Function

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First-Order Approach

Principal's canonical problem:

$$\max_{a,s(\cdot)} \int \left(\underbrace{\pi(\mathbf{x})}_{\text{Principal's value}} - \underbrace{s(\mathbf{x})}_{\text{Contract}}\right) f(\mathbf{x}|a) d\mathbf{x}$$

$$s.t. \quad (i) \ U(s(\cdot),a) \ge \overline{U}$$

$$(ii) \ a \in \arg\max_{a'} \ U(s(\cdot),a') = \int u(s(\mathbf{x})) f(\mathbf{x}|a') d\mathbf{x} - c(a') = 0$$

$$(iii) \ \underbrace{s(\mathbf{x}) \ge \underline{s}}_{\text{Contract}}$$

First-Order Approach (FOA): replace (ii) with its first-order condition (ii)'

$$\max_{a,s(\cdot)} \int \left(\underbrace{\pi(\mathbf{x})}_{\text{Principal's value}} - \underbrace{s(\mathbf{x})}_{\text{Contract}}\right) f(\mathbf{x}|a) d\mathbf{x}$$

$$s.t. \quad (i) \ U(s(\cdot), a) \geq \overline{U}$$

$$(ii)' \ U_a(s(\cdot), a) = \int u(s(\mathbf{x})) f_a(\mathbf{x}|a) d\mathbf{x} - c'(a) = 0$$

$$(iii) \ s(\mathbf{x}) \geq \underline{s}$$

Note: limited-liability $s(x) \ge \underline{s}$ for the solution existence (Mirrlees (1975)).



First-Order Approach

Optimal contract $(s^{o}(x), a^{o})$ based on the first-order approach:

$$\frac{1}{u'(s^{o}(\mathbf{x}))} = \begin{cases} \lambda + \mu \frac{f_{a}(\mathbf{x}|a^{o})}{f(\mathbf{x}|a^{o})}, & \text{if } s^{o}(\mathbf{x}) \geq \underline{s}, \\ \frac{1}{u'(\underline{s})}, & \text{otherwise}, \end{cases}$$

with $\lambda \geq 0$ and $\mu > 0$

Existence and uniqueness: Jewitt, Kadan, and Swinkels (2008)

If the agent's value function $U(s^{\circ}(\cdot), a)$,

$$U(s^{\circ}(\cdot),a) = \int u(s^{\circ}(\mathbf{x}))f(\mathbf{x}|a)d\mathbf{x} - c(a)$$

is concave in a, then the first-order approach is valid (e.g., Mirrlees (1975))



The literature

Question (Focus of the literature)

How can we make $U(s^{\circ}(\cdot), a)$ concave in a?

Strategy 1: put conditions on f(x|a), the technology:

- One-signal (i.e., x is scalar): Mirrlees (1975) and Rogerson (1985): MLRP (monotone likelihood ratio property) and CDFC (convexity of the distribution function condition)
- Multi-signal extension of CDFC: Sinclair-Desgagné (1994, GCDFC: generalized CDFC), Conlon (2009, CISP: concave increasing set property), and Jung and Kim (2015, CDFCL: convexity of the distribution function condition for the likelihood ratio)
- Too restricted (normal, gamma distributions excluded)

Strategy 2: put conditions on both u(s) and f(x|a):

- **1** Jewitt (1988) and Jung and Kim (2015)
- 2 Cannot be used with the agent's limited liability $s(x) \ge s$

Examples show the previous literature is not enough

The first-order approach cannot be justified by the previous literature in:

Example (Normal distribution)

The agent's utility is $u(s)=\frac{1}{r}s^r$, $r\leq\frac{1}{2}$, The cost function is $c(a)=D(e^{ka}-1)$, D>0, k>0, and the signal generating function has an additive form $\tilde{x}=a+\tilde{\theta}$, $\tilde{\theta}\sim N(0,\sigma^2)$ thereby

$$f(x|a) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-a)^2}{2\sigma^2}}$$

Normal distribution excluded (← its likelihood ratio unbounded)

Example (Gamma distribution)

The agent's utility is $u(s) = \frac{1}{r}s^r$, $r \leq \frac{1}{2}$, Cost function is given by c(a) = ka, k > 0, and $\tilde{x} \in (0, \infty)$ has the gamma distribution with shape parameter a, i.e.,

$$f(x|a) = \frac{x^{a-1}\beta^{-a}}{\Gamma(a)}e^{-\frac{x}{\beta}}.$$
 (1)

Gamma distribution excluded (← its likelihood ratio unbounded)



Examples show the previous literature is not enough

The first-order approach cannot be justified by the previous literature in:

Example (Exponential distribution)

The agent's utility is $u(s)=\frac{1}{r}s^r, \ r\leq \frac{1}{2}$, and cost c(a) is increasing and convex in a. The signal generating function has a multiplicative form, $\tilde{x}=h(a)\tilde{\theta}$, where $h(0)=0,\ h(a)$ is increasing and convex to a small degree, and $\tilde{\theta}$ is exponentially distributed with mean 1, i.e., the density function of $\tilde{\theta}$ is $p(\theta)=e^{-\theta},\ \theta\in [0,\infty)$. \underline{s} is low enough. Thereby

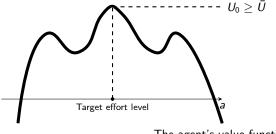
$$f(x|a) = \frac{1}{h(a)}e^{-\frac{x}{h(a)}},$$
 (2)

• A little convexity of h(a): does not satisfy Jewitt (1988) and Jung and Kim (2015)

Our paper: different approach

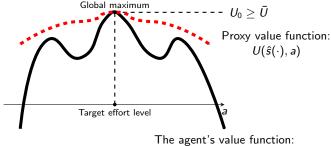
Big Question (Possibly Non-Concave Indirect Utility)

Why should the agent's value function $U(s^{\circ}(\cdot), a)$ be concave?



The agent's value function obtained from the first-order approach

Figure: Possibly Non-Concave Indirect Utility of the Agent



The agent's value function: $U(s^{\circ}(\cdot), a)$

Our approach: justify the first-order approach in all of the above examples

- Finding a proxy function $\hat{s}(\mathbf{x})$ where the proxy value $U(\hat{s}(\cdot), a)$ is maximized at $a = a^{\circ}$, the same target action level
- ② Proving $U(s^{\circ}(\cdot), a) \leq U(\hat{s}(\cdot), a)$, $\forall a$, justifying the first-order approach
- **3** A proper proxy $\hat{s}(x)$ depends on whether the limited liability binds or not

Note: impose additional conditions on the agent's cost function $c(\cdot)$

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Fundamental Lemma

Change of variables to q-space

À la Jung and Kim (2015), define the likelihood ratio

$$\tilde{q} \equiv Q_{a^o}(\tilde{\mathbf{x}}) \equiv \frac{f_a(\tilde{\mathbf{x}}|a^o)}{f(\tilde{\mathbf{x}}|a^o)}$$

The optimal contract $s^{\circ}(x)$ in q-space becomes:

$$s^{o}(x) \equiv w(q) \equiv (u')^{-1} \left(\frac{1}{\lambda + \mu q}\right)$$

The agent's indirect utility (value function) given $s^{\circ}(\cdot)$

$$u(s^{\circ}(\mathbf{x})) \equiv r(q) = \left\{ egin{array}{ll} u(w(q)) \equiv \overline{r}(q), & ext{when } q \geq q_c \\ u(\underline{s}), & ext{when } q < q_c \end{array} \right.$$

• Threshold q_c solves $u'(\underline{s})^{-1} = \lambda + \mu q_c$: limited liability starts to bind

Distribution function for q given a (possbly different from a°)

$$G(q|a) \equiv Pr[Q_{a^o}(\tilde{\mathbf{x}}) \leq q|a], \quad dG(q|a) = g(q|a)dq$$

Properties of a proxy contract

Define $U^o \geq \overline{U}$ at the optimum:

$$U^{\circ} = U(s^{\circ}(\mathbf{x}), \mathbf{a}^{\circ}) = \int s^{\circ}(\mathbf{x}) f(\mathbf{x}|\mathbf{a}^{\circ}) d\mathbf{x} - c(\mathbf{a}^{\circ})$$
(3)

Lemma (How to construct a proxy contract $\hat{s}(\cdot)$)

- (1a) $f(\mathbf{x}|a)$ satisfies that $\frac{g(q|a)}{g(q|\mathbf{a}^o)}$ is convex in $q=\frac{f_a(\mathbf{x}|\mathbf{a}^o)}{f(\mathbf{x}|\mathbf{a}^o)}$ for all a
- (2a) (DOUBLE-CROSSING PROPERTY) \exists a contract $\hat{s}(x)$ satisfying

(i)
$$\int u(\hat{s}(\mathbf{x}))f(\mathbf{x}|\mathbf{a}^{\circ})d\mathbf{x} - c(\mathbf{a}^{\circ}) = U^{\circ}$$
 (4)

(ii)
$$\int u(\hat{\mathbf{s}}(\mathbf{x}))f_{\mathbf{a}}(\mathbf{x}|\mathbf{a}^{\circ})d\mathbf{x} - c'(\mathbf{a}^{\circ}) = 0$$
 (5)

such that $\hat{r}(q) \equiv u(\hat{s}(\mathbf{x}))$ crosses $r(q) \equiv u(s^{\circ}(\mathbf{x}))$ twice starting from above

(3a) $E[\hat{r}(q)|a]$ is concave in c(a)

then using the first-order approach is justified



Intuition

(1a) and (2a) jointly imply:

$$U(s^{\circ}(\cdot),a)-U(\hat{s}(\cdot),a)=\int (r(q)-\hat{r}(q))\,g(q|a)dq\leq 0,\quad \forall a\in \mathcal{S}$$

Why?: we know that $U(s^{\circ}(\cdot), a^{\circ}) = U(\hat{s}(\cdot), a^{\circ})$ when $a = a^{\circ}$

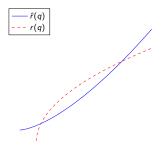


Figure: r(q) and $\hat{r}(q)$: double-crossing

As $a\uparrow$ from a': g(q|a) moves toward higher q, where $r(q) - \hat{r}(q)$ is more likely to be negative. When $a \downarrow$ from a° , the same

• (1a) condition operationalizes this intuition



(1a) and (2a) jointly imply:

$$U(s^{\circ}(\cdot),a)-U(\hat{s}(\cdot),a)=\int (r(q)-\hat{r}(q))g(q|a)dq\leq 0, \quad \forall a$$

But: It might be the following case

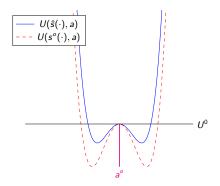


Figure: First-order approach not justified?

(3a) makes sure that $U(\hat{s}(\cdot), a)$ is maximized at $a = a^{\circ}$, therefore:

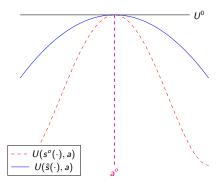


Figure: First-Order Approach Justified

So $U(s^{\circ}(\cdot), a)$ must be maximized at $a = a^{\circ}$

• The first-order approach (FOA) justified

When the Limited Liability (LL) Binds

Finding a proxy contract when (LL) binds for $q < q_c$

Define the moment generating function (MGF) of g(q|a):

$$M(a;t) \equiv \int e^{tq} g(q|a) dq.$$

Proposition (When (LL) binds for $q \leq q_c$)

Given that the likelihood ratio, $\tilde{q} \equiv \frac{f_a(\tilde{\mathbf{x}}|\mathbf{a}^o)}{f(\tilde{\mathbf{x}}|\mathbf{a}^o)}$, is unbounded below, given \mathbf{a}^o ,

- (1a) $\frac{g(q|a)}{\sigma(a|a^o)}$ is convex in $q = \frac{f_a(\mathbf{x}|a^o)}{f(\mathbf{x}|a^o)}$ for all a
- (2b) (i) there exists t > 0 such that

$$\frac{c'(a^{\circ})}{M'(a^{\circ};t)}M(a^{\circ};t)-c(a^{\circ})\leq \overline{U}-u(\underline{s})$$

and (ii) c(a) is convex in M(a;t) for such t, and

(3b) $\overline{r}(q)$ is concave in q

then the first-order approach is justified

Note: Concave $\overline{r}(q) \xrightarrow{\times}$ concave r(q) due to the kink generated by (LL)

Finding a proxy contract when (LL) binds for $q \leq q_c$

Intuition: a proxy contract $\hat{s}(\mathbf{x})$ must respect the limited liability constraint (LL). We use the following t-dependent contract

$$u(\hat{s}_t(\mathbf{x})) \equiv \hat{r}_t(q) = Ae^{tq} + B$$

which has a good property: $\hat{r}_t(q) \longrightarrow \underbrace{B \geq u(\underline{s})}_{\text{by (i) of (2b)}}$ as $q \longrightarrow -\infty$

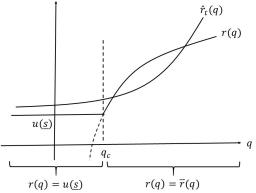


Figure: When the Limited Liability Constraint Binds for $q \leq q_c$

Finding a proxy contract when (LL) binds for $q \leq q_c$

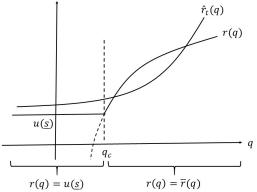


Figure: When the Limited Liability Constraint Binds for $q \leq q_c$

Note: earlier examples (cases of Normal and Gamma distributions) can be justified of their use of the first-order approach

- Both distribution features unbounded likelihood ratio (thus we need (LL))
- Jewitt (1988) and Jung and Kim (2015) assume away (LL)

When the Limited Liability (LL) Not Binds

Finding a proxy contract when (LL) does not bind

Proposition (When (LL) does not bind)

Given that the likelihood ratio, $\tilde{q} \equiv \frac{f_a(\tilde{\mathbf{x}}|a^o)}{f(\tilde{\mathbf{x}}|a^o)}$, is bounded below, given a^o ,

(1a)
$$\frac{g(q|a)}{g(q|a^o)}$$
 is convex in $q=\frac{f_a(\mathbf{x}|a^o)}{f(\mathbf{x}|a^o)}$ for all a

(2c)
$$c(a)$$
 is convex in $m(a) \equiv \int qg(q|a)dq$, and

(3c)
$$r(q) = \overline{r}(q)$$
 is concave in q

then the first-order approach is justified

Note: Now $\overline{r}(q) = r(q)$ due to the nonbinding (LL)

• In this case, finding a proxy contract $\hat{s}(x)$ is easier (no need to respect the limited liability (LL))



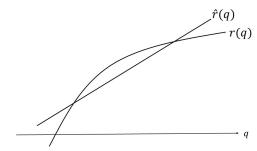


Figure: When the Limited Liability Constraint Does Not Bind

Simplest case: our proxy contract $\hat{r}(q)$ is linear in q

- (2c) makes sure under $\hat{r}(q)$, the agent will choose $a=a^{\circ}$
- (1c) and (3c) allow us to apply the lemma above (double-crossing)
- This case justifies the last example (the exponential distribution case)

Comparison with the earlier literature

To compare with Jung and Kim (2015)'s conditions (1J-1) and (1J-2):

- We introduce the total positivity of degree 3 (TP₃) (Karlin (1968))
- Our (1a) condition is related to this (TP₃) condition

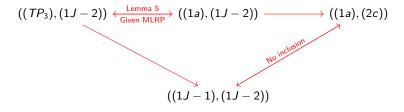


Figure: Relation Diagram between Conditions

So no direct inclusion between our paper and Jung and Kim (2015)