

Technical Appendix to Accompany
“Asymmetric Treatment of Identical Agents in Teams”

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Findings 1 – 4 below provide sufficient conditions for the values of p^A , p^B , and $p^A + p^B + \gamma p^A p^B$ that constitute the solutions to [P-S] and [P-SQ] to all lie in the $(0, 1)$ interval and for these values of p^A and p^B to be uniquely defined by the agent’s relevant first-order conditions. The proofs of these Findings proceed for the case where agent i ’s cost (k^i) can differ from agent j ’s cost (k^j). The ensuing analysis presumes that $V_F = 0$, without essential loss of generality.

To begin, define [P-S]^I to be problem [P-S] where constraints (4) and (7) in the text are replaced by:

$$[1 + \gamma p^j] T_S^i = k^i (p^i)^{\theta-1} \Rightarrow T_S^i = \frac{k^i (p^i)^{\theta-1}}{1 + \gamma p^j} \quad \text{for } j \neq i, \quad i, j \in \{A, B\}. \quad (\text{A1})$$

The solution to [P-S]^I will be the solution to [P-S] if: (i) $p^A \in (0, 1)$, $p^B \in (0, 1)$, and $p^A + p^B + \gamma p^A p^B \in (0, 1)$ at the solution to [P-S]^I; and (ii) agent i ’s choice of p^i as defined implicitly by (A1) uniquely maximizes his objective function.

Finding 1. $p^A \in (0, 1)$, $p^B \in (0, 1)$, and $p^A + p^B + \gamma p^A p^B \in (0, 1)$ at the solution to [P-S]^I if: (i) $k^A = k^B = k > V_S (2)^{\theta-1}$ when $\gamma = 0$; (ii) $[1 + \gamma]^{\frac{1}{\theta-1}} \left[2 \left(\frac{V_S}{k} \right)^{\frac{1}{\theta-1}} + \gamma [1 + \gamma]^{\frac{1}{\theta-1}} \left(\frac{V_S}{k} \right)^{\frac{2}{\theta-1}} \right] < 1$ when $\gamma > 0$; and (iii) $k > V_S (2)^{\theta-1}$ when $\gamma \in (-1, 0)$.

Proof. Given p^j , agent i chooses p^i to maximize:

$$\Pi^i(p^i | p^j) = p(p^i, p^j) T_S^i - \frac{k^i}{\theta} (p^i)^\theta. \quad (\text{A2})$$

Differentiating (A2) provides:

$$\frac{\partial \Pi^i(\cdot)}{\partial p^i} = \frac{\partial p(\cdot)}{\partial p^i} T_S^i - k^i (p^i)^{\theta-1}. \quad (\text{A3})$$

Equation (1) in the text implies $\frac{\partial p(\cdot)}{\partial p^i} \in \{0, 1 + \gamma p^j\}$ for $\tilde{p} \neq 1$, where $\tilde{p} \equiv p^A + p^B + \gamma p^A p^B$. Consequently, (A3) implies that for $\tilde{p} < 1$:

$$\frac{\partial^2 \Pi^i(\cdot)}{\partial (p^i)^2} = -k^i [\theta - 1] (p^i)^{\theta-2} < 0. \quad (\text{A4})$$

Furthermore, from (A3):

$$\frac{\partial \Pi^i(\cdot)}{\partial p^i} = \begin{cases} [1 + \gamma p^j] T_S^i - k^i (p^i)^{\theta-1} & \text{when } \tilde{p} \in (0, 1) \\ -k^i (p^i)^{\theta-1} & \text{when } \tilde{p} > 1. \end{cases} \quad (\text{A5})$$

Equation (1) in the text and (A5) imply:

$$p^i(p^j) = \begin{cases} \left(\frac{[1+\gamma p^j]T_S^i}{k^i}\right)^{\frac{1}{\theta-1}} & \text{when } \tilde{p} \in (0, 1) \\ 1 - p^j & \text{when } \tilde{p} > 1. \end{cases} \quad (\text{A6})$$

Case 1. $\gamma = 0$.

(A6) implies that $p^A < 1$, $p^B < 1$, and $p^A + p^B < 1$ in this case if $\left[\frac{T_S^i}{k^i}\right]^{\frac{1}{\theta-1}} < \frac{1}{2}$ for $i = A, B$. Therefore, because the principal will never pay an agent more than the value of success (V_S), a sufficient condition for unique, interior values of p^A and p^B is:

$$\left[\frac{V_S}{k^i}\right]^{\frac{1}{\theta-1}} < \frac{1}{2} \Leftrightarrow 2[V_S]^{\frac{1}{\theta-1}} < [k^i]^{\frac{1}{\theta-1}} \Leftrightarrow k^i > V_S(2)^{\theta-1} \text{ for } i = A, B.$$

Case 2. $\gamma > 0$.

(A6) implies that if $p^A + p^B + \gamma p^A p^B \in (0, 1)$, then :

$$\frac{\partial p^i(\cdot)}{\partial p^j} = \left(\frac{T_S^i}{k^i}\right)^{\frac{1}{\theta-1}} \left[\frac{\gamma}{\theta-1}\right] [1 + \gamma p^j]^{\frac{2-\theta}{\theta-1}} > 0, \quad \text{and} \quad (\text{A7})$$

$$\frac{\partial^2 p^i(\cdot)}{\partial (p^j)^2} = - \left(\frac{T_S^i}{k^i}\right)^{\frac{1}{\theta-1}} \left[\frac{\theta-2}{[\theta-1]^2}\right] \gamma^2 [1 + \gamma p^j]^{\frac{3-2\theta}{\theta-1}} < 0. \quad (\text{A8})$$

(A7) and (A8) imply that as long as $p(\cdot) \in (0, 1)$, agent i 's best response function is an increasing, concave function of p^j . Also, (A6) implies that $p^i = \left[\frac{T_S^i}{k^i}\right]^{\frac{1}{\theta-1}}$ when $p^j = 0$.

We seek conditions sufficient to ensure the best response functions of agents A and B intersect at a point that lies (strictly) within the region in the (p^A, p^B) -plane bounded by the $p^A = 0$ axis, the $p^B = 0$ axis, and the curve $p^A + p^B + \gamma p^A p^B = 1$. Formally, we seek conditions that ensure the solutions to (A9) and (A10) also satisfy (A11).

$$[1 + \gamma p^B] T_S^A - k^A (p^A)^{\theta-1} = 0. \quad (\text{A9})$$

$$[1 + \gamma p^A] T_S^B - k^B (p^B)^{\theta-1} = 0. \quad (\text{A10})$$

$$p^A + p^B + \gamma p^A p^B \in (0, 1). \quad (\text{A11})$$

When $k^A = k^B = k$, it suffices to identify conditions that ensure the solutions to (A12) and (A13) satisfy (A11).

$$[1 + \gamma p^B] V_S - k (p^A)^{\theta-1} = 0. \quad (\text{A12})$$

$$[1 + \gamma p^A] V_S - k (p^B)^{\theta-1} = 0. \quad (\text{A13})$$

This is the case because $T^A \leq V_S$ and $T^B \leq V_S$ at the solution to [P-S] and because, from (A9) and (A10):

$$\frac{dp^i}{dT_S^i} = \frac{1 + \gamma p^j}{k^i[\theta-1](p^i)^{\theta-2}} > 0 \text{ when } \gamma > 0. \quad (\text{A14})$$

Therefore, the values of p^A and p^B that solve (A9) and (A10) will be less than the corresponding values of p^A and p^B that solve (A12) and (A13) when $\gamma > 0$.

(A9) implies:

$$\begin{aligned} k^A (p^A)^{\theta-1} &= [1 + \gamma p^B] T_S^A \leq [1 + \gamma] T_S^A \\ \Rightarrow (p^A)^{\theta-1} &\leq \frac{[1 + \gamma] T_S^A}{k^A} \Rightarrow p^A \leq \left(\frac{[1 + \gamma] T_S^A}{k^A} \right)^{\frac{1}{\theta-1}}. \end{aligned} \quad (\text{A15})$$

Similarly, (A10) implies:

$$p^B \leq \left(\frac{[1 + \gamma] T_S^B}{k^B} \right)^{\frac{1}{\theta-1}}. \quad (\text{A16})$$

(A15) and (A16) imply:

$$\begin{aligned} p^A + p^B + \gamma p^A p^B &\leq \left(\frac{[1 + \gamma] T_S^A}{k^A} \right)^{\frac{1}{\theta-1}} + \left(\frac{[1 + \gamma] T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} + \gamma \left(\frac{[1 + \gamma] T_S^A}{k^A} \right)^{\frac{1}{\theta-1}} \left(\frac{[1 + \gamma] T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} \\ &= [1 + \gamma]^{\frac{1}{\theta-1}} \left[\left(\frac{T_S^A}{k^A} \right)^{\frac{1}{\theta-1}} + \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} + \gamma [1 + \gamma]^{\frac{1}{\theta-1}} \left(\frac{T_S^A}{k^A} \right)^{\frac{1}{\theta-1}} \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} \right] \\ &\leq [1 + \gamma]^{\frac{1}{\theta-1}} \left[\left(\frac{V_S}{k^A} \right)^{\frac{1}{\theta-1}} + \left(\frac{V_S}{k^B} \right)^{\frac{1}{\theta-1}} + \gamma [1 + \gamma]^{\frac{1}{\theta-1}} \left(\frac{V_S}{k^A} \right)^{\frac{1}{\theta-1}} \left(\frac{V_S}{k^B} \right)^{\frac{1}{\theta-1}} \right]. \end{aligned} \quad (\text{A17})$$

(A15) – (A17) imply that if $k^A = k^B = k$ and if $[1 + \gamma]^{\frac{1}{\theta-1}} \left[2 \left(\frac{V_S}{k} \right)^{\frac{1}{\theta-1}} + \gamma [1 + \gamma]^{\frac{1}{\theta-1}} \left(\frac{V_S}{k} \right)^{\frac{2}{\theta-1}} \right] < 1$, then $p^A < 1$, $p^B < 1$ and $p^A + p^B + \gamma p^A p^B < 1$ at the solution to [P-S]^I.

Case 3. $\gamma < 0$.

(A7) and (A8) imply that as long as $p \in (0, 1)$, agent i 's optimal choice of p^i is a decreasing, convex function of p^j . Also, (A6) implies that $p^i = \left[\frac{T_S^i}{k^i} \right]^{\frac{1}{\theta-1}}$ when $p^j = 0$. From (A14), if $|\gamma| < 1$, then $\frac{dp^i}{dT_S^i} > 0$ when $p^j < 1$. Consequently, it suffices to ensure that the solutions to (A12) and (A13) satisfy (A11). (A9), (A10), (A12), and (A13) imply:

$$p^i = \left(\frac{[1 + \gamma p^j] V_S}{k^i} \right)^{\frac{1}{\theta-1}} < \left(\frac{V_S}{k^i} \right)^{\frac{1}{\theta-1}} < \frac{1}{2} \text{ for } i \in \{A, B\} \text{ when } k^i > V_S (2)^{\theta-1}.$$

Therefore, $p^A < 1$, $p^B < 1$, and $p^A + p^B + \gamma p^A p^B < 1$ when $k^i > V_S (2)^{\theta-1}$ for $i = A, B$. Also, if $|\gamma| < 1$, then (A9) and (A10) imply $p^A > 0$ and $p^B > 0$. Furthermore, $p^A + p^B + \gamma p^A p^B = p^A [1 + \gamma p^B] + p^B > 0$ when $|\gamma| < 1$. ■

Finding 2. Agent $i \in \{A, B\}$'s choice of p^i as defined by (A1) uniquely maximizes his objective function when the conditions identified in Finding 1 hold.

Proof. From (A4), $\Pi^i(p^i | p^j)$ is a strictly concave function of p^i for $p^j \in (0, 1)$ for $j \neq i$,

$i, j \in \{A, B\}$. Therefore, when the conditions identified in Finding 1 hold, p^i as defined by (A1) uniquely maximizes (A2). ■

Now define [P-SQ]^J to be problem [P-SQ] where constraint (4) in the text is replaced by (A1) for $i = B$ and where constraint (5) in the text is replaced by:

$$T_S^A = \frac{k^A (p^A)^{\theta-1}}{1 + \gamma p^B + \frac{\gamma p^B}{\theta-1}} = \frac{k^A (p^A)^{\theta-1}}{1 + \left[\frac{\theta}{\theta-1}\right] \gamma p^B}. \quad (\text{A18})$$

(A18) is derived by first differentiating (A1) when $i = B$ to obtain:

$$\gamma T_S^B dp^A = k^B [\theta - 1] (p^B)^{\theta-2} dp^B \Rightarrow \frac{dp^B}{dp^A} = \frac{\gamma T_S^B}{k^B [\theta - 1] (p^B)^{\theta-2}}. \quad (\text{A19})$$

Then, differentiating constraint (5) in the text (with $k = k^A$) reveals that for $p(\cdot) \in (0, 1)$:

$$\left[1 + \gamma p^B + (1 + \gamma p^A) \frac{dp^B}{dp^A}\right] T_S^A = k^A (p^A)^{\theta-1} \Rightarrow T_S^A = \frac{k^A (p^A)^{\theta-1}}{1 + \gamma p^B + [1 + \gamma p^A] \frac{dp^B}{dp^A}}. \quad (\text{A20})$$

(A1) and (A19) imply:

$$[1 + \gamma p^A] \frac{dp^B}{dp^A} = \frac{[1 + \gamma p^A] \gamma T_S^B}{k^B [\theta - 1] (p^B)^{\theta-2}} = \frac{\gamma k^B (p^B)^{\theta-1}}{k^B [\theta - 1] (p^B)^{\theta-2}} = \frac{\gamma p^B}{\theta - 1}. \quad (\text{A21})$$

Substituting (A21) into (A20) provides (A18).

Finding 3. $p^A \in (0, 1)$, $p^B \in (0, 1)$, and $p^A + p^B + \gamma p^A p^B \in (0, 1)$ at the solution to [P-SQ]^J if: (i) $k^A = k^B = k > V_S (2)^{\theta-1}$ when $\gamma = 0$; (ii) $k > \tilde{k}$ when $\gamma > 0$, where \tilde{k} is defined by $(\tilde{p}^A)^{\theta-1} = \frac{V_S}{k} + \left[\frac{\gamma(1+\gamma)\theta}{\theta-1}\right] (V_S)^{\frac{\theta}{\theta-1}} \left(\frac{1}{k}\right)^{\frac{\theta}{\theta-1}}$ and \tilde{p}^A is defined by $\tilde{p}^A + [1 + \gamma \tilde{p}^A] \left(\frac{[1+\gamma\tilde{p}^A]V_S}{k}\right)^{\frac{1}{\theta-1}} = 1$; and (iii) $k > V_S (2)^{\theta-1}$ when $\gamma \in (-1, 0)$.

Proof.

Case 1. $\gamma = 0$.

The analysis in this case is identical to the analysis in the proof of Finding 1.

Case 2. $\gamma > 0$.

The analysis for agent B coincides with the corresponding analysis in the proof of Finding 1.

Agent A will pick his preferred point on agent B's reaction function as defined by (A6) for $i = B$. Let $\pi^A(p^A) \equiv \Pi^A(p^A | p^B(p^A))$ denote agent A's expected profit in the present setting. Also let $\hat{p}^A \leq 1$ denote the smallest value of p^A on the portion of agent B's reaction function defined by the equation $p^A + p^B + \gamma p^A p^B = 1$.

Agent A will never choose to deliver a contribution in excess of \widehat{p}^A because he can secure the same aggregate probability of success ($p(\cdot) = 1$) at lower personal cost by setting $p^A = \widehat{p}^A$. Therefore, conditions that ensure $\left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=0} > 0$ and $\left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=\widehat{p}^A} < 0$ are sufficient to ensure agent A will choose a $p^A \in (0, 1)$.

From (A20), for $p^A + p^B + \gamma p^A p^B \in (0, 1)$:

$$\frac{d\pi^A(\cdot)}{dp^A} = [1 + \gamma p^B] T_S^A + [1 + \gamma p^A] T_S^A \frac{dp^B}{dp^A} - k^A (p^A)^{\theta-1}. \quad (\text{A22})$$

Solving (A10) for p^B provides:

$$p^B = \left(\frac{[1 + \gamma p^A] T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} \Rightarrow \frac{dp^B}{dp^A} = \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} \left[\frac{\gamma}{\theta-1} \right] [1 + \gamma p^A]^{\frac{2-\theta}{\theta-1}}. \quad (\text{A23})$$

Substituting from (A23) into (A22) provides:

$$\begin{aligned} \frac{d\pi^A(\cdot)}{dp^A} &= T_S^A + \gamma T_S^A \left(\frac{[1 + \gamma p^A] T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} \\ &\quad + [1 + \gamma p^A] T_S^A \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} \left[\frac{\gamma}{\theta-1} \right] [1 + \gamma p^A]^{\frac{2-\theta}{\theta-1}} - k^A (p^A)^{\theta-1} \\ &= T_S^A + \gamma \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} [1 + \gamma p^A]^{\frac{1}{\theta-1}} T_S^A \left[\frac{\theta}{\theta-1} \right] - k^A (p^A)^{\theta-1}. \end{aligned} \quad (\text{A24})$$

(A24) implies:

$$\left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=0} = T_S^A + \gamma \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} T_S^A \left[\frac{\theta}{\theta-1} \right] > 0; \quad \text{and} \quad (\text{A25})$$

$$\begin{aligned} \left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=\widehat{p}^A} &= T_S^A + \gamma \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} [1 + \gamma \widehat{p}^A]^{\frac{1}{\theta-1}} T_S^A \left[\frac{\theta}{\theta-1} \right] - k^A (\widehat{p}^A)^{\theta-1} \\ &< V_S + \gamma \left(\frac{V_S}{k^B} \right)^{\frac{1}{\theta-1}} [1 + \gamma \widehat{p}^A]^{\frac{1}{\theta-1}} V_S \left[\frac{\theta}{\theta-1} \right] - k^A (\widehat{p}^A)^{\theta-1}. \end{aligned} \quad (\text{A26})$$

The inequality in (A26) holds because $T_S^A \leq V_S$ and $T_S^B \leq V_S$ at the solution to [P-SQ]^I. (A26) implies:

$$\left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=\widehat{p}^A} < 0 \quad \text{if} \quad k^A (\widehat{p}^A)^{\theta-1} \geq V_S + \left[\frac{\gamma \theta}{\theta-1} \right] \left(\frac{V_S}{k^B} \right)^{\frac{1}{\theta-1}} [1 + \gamma \widehat{p}^A]^{\frac{1}{\theta-1}} V_S. \quad (\text{A27})$$

Since $\widehat{p}^A \leq 1$, the weak inequality in (A27) will hold when $k^A = k^B = k$ if:

$$(\widehat{p}^A)^{\theta-1} \geq \frac{V_S}{k} + \left[\frac{\gamma (1 + \gamma) \theta}{\theta - 1} \right] (V_S)^{\frac{\theta}{\theta-1}} \left(\frac{1}{k} \right)^{\frac{\theta}{\theta-1}}. \quad (\text{A28})$$

From equation (1) in the text and (A23), \widehat{p}^A is the solution to:

$$\widehat{p}^A + p^B + \gamma \widehat{p}^A p^B = 1 \quad \text{and} \quad p^B = \left(\frac{[1 + \gamma \widehat{p}^A] T^B}{k^B} \right)^{\frac{1}{\theta-1}}. \quad (\text{A29})$$

(A29) implies that \widehat{p}^A is the solution to:

$$\widehat{p}^A + \left(\frac{[1 + \gamma \widehat{p}^A] T^B}{k^B} \right)^{\frac{1}{\theta-1}} + \gamma \widehat{p}^A \left(\frac{[1 + \gamma \widehat{p}^A] T^B}{k^B} \right)^{\frac{1}{\theta-1}} = 1. \quad (\text{A30})$$

Let \widetilde{p}^A be the solution to:

$$\widetilde{p}^A + \left(\frac{[1 + \gamma \widetilde{p}^A] V_S}{k^B} \right)^{\frac{1}{\theta-1}} + \gamma \widetilde{p}^A \left(\frac{[1 + \gamma \widetilde{p}^A] V_S}{k^B} \right)^{\frac{1}{\theta-1}} = 1. \quad (\text{A31})$$

Since $V_S \geq T^B$, (A30) and (A31) imply that $\widetilde{p}^A \leq \widehat{p}^A$ for given $\gamma > 0$ and $k^B > 0$. Therefore, when $k^A = k^B = k$, the inequality in (A28) holds if:

$$(\widetilde{p}^A)^{\theta-1} \geq \frac{V_S}{k} + \left[\frac{\gamma (1 + \gamma) \theta}{\theta - 1} \right] (V_S)^{\frac{\theta}{\theta-1}} \left(\frac{1}{k} \right)^{\frac{\theta}{\theta-1}}. \quad (\text{A32})$$

Define \widetilde{k} by:

$$(\widetilde{p}^A)^{\theta-1} = \frac{V_S}{\widetilde{k}} + \left[\frac{\gamma (1 + \gamma) \theta}{\theta - 1} \right] (V_S)^{\frac{\theta}{\theta-1}} \left(\frac{1}{\widetilde{k}} \right)^{\frac{\theta}{\theta-1}}. \quad (\text{A33})$$

(A31) implies that \widetilde{p}^A is an increasing function of $k^A = k^B = k$. Furthermore, the expression to the right of the inequality in (A32) is a decreasing function of k . Therefore, as k is increased above \widetilde{k} , the inequality in (A32) will continue to hold, and so $\left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=\widetilde{p}^A} < 0$ for $k > \widetilde{k}$.

Case 3. $\gamma < 0$.

From (A25):

$$\begin{aligned} \left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=0} &= T^A \left[1 + \gamma \left(\frac{T^B}{k^B} \right)^{\frac{1}{\theta-1}} \left(\frac{\theta}{\theta-1} \right) \right] \\ &\geq T^A \left[1 + \gamma \left(\frac{V_S}{k^B} \right)^{\frac{1}{\theta-1}} \left(\frac{\theta}{\theta-1} \right) \right] > T^A \left[1 + \frac{\gamma}{2} \left(\frac{\theta}{\theta-1} \right) \right]. \end{aligned} \quad (\text{A34})$$

The strict inequality in (A34) holds because $\gamma < 0$ and because $k^i > V_S (2)^{\theta-1}$ for $i = A, B$ implies $\left(\frac{V_S}{k^B} \right)^{\frac{1}{\theta-1}} < \frac{1}{2}$. (A34) implies:

$$\left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=0} > 0 \quad \text{if} \quad 1 + \frac{\gamma}{2} \left[\frac{\theta}{\theta-1} \right] > 0 \quad \Leftrightarrow \quad \gamma > -\frac{2[\theta-1]}{\theta}. \quad (\text{A35})$$

(A35) implies that $\left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=0} > 0$ if $\gamma \in (-\frac{2[\theta-1]}{\theta}, 0)$ when $k^A = k^B = k > V_S (2)^{\theta-1}$.

From (A23), $\gamma > -1$ ensures $1 + \gamma p^A > 0$, and so ensures $p^B > 0$. Also, since $\theta > 2$, $\frac{\theta-1}{\theta} > \frac{1}{2}$, which implies $\frac{2[\theta-1]}{\theta} > 1$. Therefore, if $|\gamma| < 1$, then $|\gamma| < \frac{2[\theta-1]}{\theta}$. Consequently, $\gamma > -1$ ensures $\gamma > -\frac{2[\theta-1]}{\theta}$ when $\gamma < 0$. Therefore, $\gamma \in (-1, 0)$ ensures $\left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=0} > 0$ and also ensures $p^B > 0$.

To derive conditions that ensure $\left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=\hat{p}^A} < 0$, let $\hat{p}_{\gamma < 0}^A$ denote the value of \hat{p}^A for given $\gamma < 0$. (A24) implies:

$$\begin{aligned} \left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=\hat{p}_{\gamma < 0}^A} &= T_S^A + \gamma \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} [1 + \gamma \hat{p}_{\gamma < 0}^A]^{\frac{1}{\theta-1}} T_S^A \left[\frac{\theta}{\theta-1} \right] - k^A (\hat{p}_{\gamma < 0}^A)^{\theta-1} \\ &< T_S^A - k^A (\hat{p}_{\gamma < 0}^A)^{\theta-1} \quad \text{since } \gamma \in (-1, 0) \\ &= k^A \left[\frac{T_S^A}{k^A} - (\hat{p}_{\gamma < 0}^A)^{\theta-1} \right] \leq k^A \left[\frac{V_S}{k^A} - (\hat{p}_{\gamma < 0}^A)^{\theta-1} \right] \\ \Rightarrow \left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A=\hat{p}_{\gamma < 0}^A} &< 0 \text{ if } \frac{V_S}{k^A} - (\hat{p}_{\gamma < 0}^A)^{\theta-1} < 0 \Leftrightarrow \hat{p}_{\gamma < 0}^A > \left(\frac{V_S}{k^A} \right)^{\frac{1}{\theta-1}}. \end{aligned} \quad (\text{A36})$$

The first equality in (A29) implies that $\hat{p}_{\gamma < 0}^A > \hat{p}_{\gamma=0}^A$ for a fixed \hat{p}^B . The second equality in (A29) implies that \hat{p}^B decreases as \hat{p}^A increases when $\gamma < 0$. Therefore, $\hat{p}_{\gamma < 0}^A > \hat{p}_{\gamma=0}^A$. Consequently:

$$\hat{p}_{\gamma=0}^A > \left(\frac{V_S}{k^A} \right)^{\frac{1}{\theta-1}} \Rightarrow \hat{p}_{\gamma < 0}^A > \left(\frac{V_S}{k^A} \right)^{\frac{1}{\theta-1}} \Rightarrow \frac{V_S}{k^A} - (\hat{p}_{\gamma < 0}^A)^{\theta-1} < 0. \quad (\text{A37})$$

(A29) also implies:

$$\hat{p}_{\gamma=0}^B = \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}}, \quad \text{and so } \hat{p}_{\gamma=0}^A = 1 - \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}}. \quad (\text{A38})$$

(A37) and (A38) imply:

$$1 - \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} > \left(\frac{V_S}{k^A} \right)^{\frac{1}{\theta-1}} \Rightarrow \frac{V_S}{k^A} - (\hat{p}_{\gamma < 0}^A)^{\theta-1} < 0. \quad (\text{A39})$$

$1 - \left(\frac{T_S^B}{k^B} \right)^{\frac{1}{\theta-1}} \geq 1 - \left(\frac{V_S}{k^B} \right)^{\frac{1}{\theta-1}}$ since $T_S^B \leq V_S$. Therefore, from (A39):

$$1 - \left(\frac{V_S}{k^B} \right)^{\frac{1}{\theta-1}} > \left(\frac{V_S}{k^A} \right)^{\frac{1}{\theta-1}} \Rightarrow \frac{V_S}{k^A} - (\hat{p}_{\gamma < 0}^A)^{\theta-1} < 0. \quad (\text{A40})$$

(A40) implies:

$$\left(\frac{V_S}{k^B}\right)^{\frac{1}{\theta-1}} + \left(\frac{V_S}{k^A}\right)^{\frac{1}{\theta-1}} < 1 \Rightarrow \frac{V_S}{k^A} - (\widehat{p}_{\gamma < 0}^A)^{\theta-1} < 0. \quad (\text{A41})$$

(A36) and (A41) imply:

$$\left(\frac{V_S}{k^A}\right)^{\frac{1}{\theta-1}} + \left(\frac{V_S}{k^B}\right)^{\frac{1}{\theta-1}} < 1 \Rightarrow \left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A = \widehat{p}_{\gamma < 0}^A} < 0. \quad (\text{A42})$$

Therefore, $\left. \frac{d\pi^A(\cdot)}{dp^A} \right|_{p^A = \widehat{p}_{\gamma < 0}^A} < 0$ if $k^i > V_S (2)^{\theta-1} \Leftrightarrow \left(\frac{V_S}{k^i}\right)^{\frac{1}{\theta-1}} < \frac{1}{2}$ for $i = A, B$. ■

Finding 4. Suppose the conditions identified in Finding 3 hold. Further suppose $|\gamma|$ is sufficiently small when $\gamma < 0$. Then agent B's choice of p^B as defined by (A1) with $i = B$ uniquely maximizes his objective function, and agent A's choice of p^A as defined by (A18) uniquely maximizes his objective function.

Proof. From (A4), $\Pi^B(p^B|p^A)$ is a strictly concave function of p^B . Therefore, the value of p^B identified in (A1) with $i = B$ uniquely maximizes $\Pi^B(\cdot)$ when the conditions identified in Finding 3 hold.

Differentiating (A24) provides:

$$\frac{d^2\pi^A(\cdot)}{d(p^A)^2} = \left(\frac{T_S^B}{k^B}\right)^{\frac{1}{\theta-1}} T_S^A \left[\frac{\gamma^2 \theta}{(\theta-1)^2} \right] [1 + \gamma p^A]^{\frac{2-\theta}{\theta-1}} - k^A [\theta-1] (p^A)^{\theta-2}. \quad (\text{A43})$$

Differentiating (A43) provides:

$$\frac{d^3\pi^A(\cdot)}{d(p^A)^3} = - \left(\frac{T_S^B}{k^B}\right)^{\frac{1}{\theta-1}} T_S^A \left[\frac{\gamma^3 \theta (\theta-2)}{(\theta-1)^3} \right] [1 + \gamma p^A]^{\frac{3-2\theta}{\theta-1}} - k^A [\theta-1] [\theta-2] (p^A)^{\theta-3}. \quad (\text{A44})$$

(A44) implies that $\frac{d^3\pi^A(\cdot)}{d(p^A)^3} < 0$ when $\gamma > 0$, since $\theta > 2$.

(A24) and (A43) imply that $\pi^A(\cdot)$ is increasing and convex at $p^A = 0$ when $\gamma > 0$. Furthermore, $\pi^A(\cdot)$ is decreasing at \widehat{p}^A when $\gamma > 0$ and the conditions in Finding 3 hold. Therefore, since $\frac{d^3\pi^A(\cdot)}{d(p^A)^3} < 0$ when $\gamma > 0$, the value of $p^A \in (0, \widehat{p}^A)$ defined by (A18) uniquely maximizes agent A's expected profit.

It is also apparent from (A44) that $\frac{d^2\pi^A(\cdot)}{d(p^A)^2}$ is a continuous, decreasing function of p^A when $\gamma = 0$. Therefore, if $\pi^A(\cdot)$ is increasing at $p^A = 0$ and decreasing at \widehat{p}^A when $\gamma = 0$, then there will exist a $\widetilde{\gamma} < 0$ such that for all $\gamma \in [\widetilde{\gamma}, 0)$, the value of $p^A \in (0, \widehat{p}^A)$ at which the expression in (A24) is zero uniquely maximizes agent A's expected profit. Consequently, agent A's choice of p^A as defined by (A18) uniquely maximizes his objective function when the conditions identified in Finding 3 hold and when $\gamma \in [\widetilde{\gamma}, 0)$. ■