

Online Appendix to “Dynamic Coordination with Informational Externalities”

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OA.1 Omitted Proofs for **Section 3**

OA.1.1 Proof of **Lemma 6**

Leader x 's expected payoff from stopping at t is

$$\mathcal{L}(x, t) = \lim_{\varepsilon \rightarrow 0} \left(q_L(x) \int_0^{t-\varepsilon} e^{-r\tau} dG_F^1(\tau) H - (1 - q_L(x)) \int_0^{t-\varepsilon} e^{-r\tau} dG_F^0(\tau) L \right).$$

Follower y 's expected payoff from stopping at t is

$$\begin{aligned} \mathcal{F}(y, t) = & e^{-rt} \left(q_F(y) \left((1 - G_L^1(t))H + (1 - G_L^0(t))L \right) - (1 - G_L^0(t))L \right) \\ & - e^{-rt} \lim_{\varepsilon \rightarrow 0} \left(q_F(y) (1 - G_L^1(t - \varepsilon)) + (1 - q_F(y)) (1 - G_L^0(t - \varepsilon)) \right) c. \end{aligned}$$

I show the leader's expected payoff is supermodular and the follower's is submodular.

Denote $\Delta\mathcal{L}(x, t, t') = \mathcal{L}(x, t') - \mathcal{L}(x, t)$. For $t' > t$ and $x' > x$,

$$\begin{aligned} & \Delta\mathcal{L}(x', t, t') - \Delta\mathcal{L}(x, t, t') \\ = & \lim_{\varepsilon \rightarrow 0} (q_L(x') - q_L(x)) \left(\int_{t-\varepsilon}^{t'-\varepsilon} e^{-r\tau} dG_F^1(\tau) H + \int_{t-\varepsilon}^{t'-\varepsilon} e^{-r\tau} dG_F^0(\tau) L \right). \end{aligned}$$

By MLRP, $q_L(x') - q_L(x) > 0$. For $t' > t$, $G_F^\theta(t') \geq G_F^\theta(t)$. So $\Delta\mathcal{L}(x', t, t') - \Delta\mathcal{L}(x, t, t') > 0$. Therefore, $\mathcal{L}(x, t)$ is supermodular in (x, t) . By Topkis's theorem, $\sigma_L(x) = \arg \max_{t \geq 0} \mathcal{L}(x, t)$ is non-decreasing in x .

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Denote $\Delta\mathcal{F}(y, t, t') = \mathcal{F}(y, t') - \mathcal{F}(y, t)$. For $t' > t$ and $y' > y$,

$$\begin{aligned} & \Delta\mathcal{F}(y', t, t') - \Delta\mathcal{F}(y, t, t') \\ &= (q_F(y') - q_F(y)) \left(e^{-rt'}(1 - G_L^1(t')) - e^{-rt}(1 - G_L^1(t)) \right) H \\ & \quad - (q_F(y') - q_F(y)) \left(e^{-rt}(1 - G_L^0(t)) - e^{-rt'}(1 - G_L^0(t')) \right) L \\ & \quad - \lim_{\varepsilon \rightarrow 0} c \left(e^{-r(t'-\varepsilon)}(q_F(y') - q_F(y)) \left((1 - G_L^1(t' - \varepsilon)) + (1 - G_L^0(t' - \varepsilon)) \right) \right. \\ & \quad \left. + e^{-r(t-\varepsilon)}(q_F(y') - q_F(y)) \left((1 - G_L^1(t - \varepsilon)) + (1 - G_L^0(t - \varepsilon)) \right) \right). \end{aligned}$$

By MLRP, $q_F(y') - q_F(y) > 0$. For $t' > t$, $e^{-rt'}(1 - G_L^\theta(t')) < e^{-rt}(1 - G_L^\theta(t)) \leq e^{-rt}(1 - G_L^\theta(t))$. So $\Delta\mathcal{F}(y', t, t') - \Delta\mathcal{F}(y, t, t') < 0$. Therefore, $\mathcal{F}(y, t)$ is submodular in (y, t) . By Topkis's theorem, $\sigma_F(y) = \arg \max_{t \geq 0} \mathcal{F}(y, t)$ is non-increasing in y .

OA.2 Omitted Proofs for **Section 4**

OA.2.1 Proof of **Lemma 10**

Define $Q^\theta(\mu) := (1 - F^\theta(\mu))/(1 - \hat{F}^\theta(\mu))$. It follows directly from (6) that $h(\mu) > \hat{h}(\mu)$ if and only if $Q^1(\mu) < Q^0(\mu)$. Moreover, by (6), for all $\mu \in (0, 1)$,

$$\frac{f^0(\mu)}{\hat{f}^0(\mu)} = \frac{f^1(\mu)}{\hat{f}^1(\mu)} = \frac{f^0(\mu) + f^1(\mu)}{\hat{f}^0(\mu) + \hat{f}^1(\mu)}. \quad (\text{OA.1})$$

Because $F \succ_{\text{ULR}} \hat{F}$, all three ratios in (OA.1) are unimodal and symmetric about $1/2$. Then $Q^\theta(\mu)$ is unimodal with maximum achieved at $\hat{\mu}_Q^\theta < 1/2$ (Hopkins and Kornienko, 2007, Proposition 2). Moreover, $\lim_{\mu \rightarrow 0} Q^1(\mu) = \lim_{\mu \rightarrow 0} Q^0(\mu) = 1$ and

$$\lim_{\mu \rightarrow 1} Q^1(\mu) = \lim_{\mu \rightarrow 1} \frac{1 - F^1(\mu)}{1 - \hat{F}^1(\mu)} = \lim_{\mu \rightarrow 1} \frac{f^1(\mu)}{\hat{f}^1(\mu)} = \lim_{\mu \rightarrow 1} \frac{f^0(\mu)}{\hat{f}^0(\mu)} = \lim_{\mu \rightarrow 1} \frac{1 - F^0(\mu)}{1 - \hat{F}^0(\mu)} = \lim_{\mu \rightarrow 1} Q^0(\mu).$$

The proof concerns comparing the derivatives of Q^1 and Q^0 , which are given by

$$\frac{dQ^1}{d\mu} = \frac{f^1(\mu)}{1 - \hat{F}^1(\mu)} \left(Q^1(\mu) - \frac{f^1(\mu)}{\hat{f}^1(\mu)} \right) \text{ and } \frac{dQ^0}{d\mu} = \frac{f^0(\mu)}{1 - \hat{F}^0(\mu)} \left(Q^0(\mu) - \frac{f^0(\mu)}{\hat{f}^0(\mu)} \right).$$

By MLRP and (OA.1), $f^1(\mu)/(1 - \hat{F}^1(\mu)) < f^0(\mu)/(1 - \hat{F}^0(\mu))$ for all μ .

Consider $\mu \geq \max\{\hat{\mu}_Q^1, \hat{\mu}_Q^0\}$, then both $Q^1(\mu)$ and $Q^0(\mu)$ are decreasing. Suppose there exists $\tilde{\mu}$ such that $Q^0(\tilde{\mu}) \leq Q^1(\tilde{\mu})$. Then at $\tilde{\mu}$, $dQ^0/d\mu < dQ^1/d\mu < 0$. This is a contradiction because $\lim_{\mu \rightarrow 1} Q^1(\mu) = \lim_{\mu \rightarrow 1} Q^0(\mu)$.

At $\mu = \max\{\hat{\mu}_Q^1, \hat{\mu}_Q^0\}$, one of $dQ^1/d\mu$ and $dQ^0/d\mu$ is zero and the other is strictly negative. As is shown above, $Q^1(\mu) < Q^0(\mu)$, so it must be that $dQ^1/d\mu < 0$ and $dQ^0/d\mu = 0$. This implies $\hat{\mu}_Q^1 < \hat{\mu}_Q^0$.

Consider $\mu \in (\hat{\mu}_Q^1, \hat{\mu}_Q^0)$, then Q^1 is decreasing and Q^0 is increasing. $dQ^1/d\mu < 0$ and $dQ^0/d\mu > 0$ implies $Q^1(\mu) < f^1(\mu)/\hat{f}^1(\mu) = f^0(\mu)/\hat{f}^0(\mu) < Q^0(\mu)$.

Consider $\mu \leq \hat{\mu}_Q^1$, then both $Q^1(\mu)$ and $Q^0(\mu)$ are increasing. Suppose there exists $\tilde{\mu}$ such that $Q^0(\tilde{\mu}) \leq Q^1(\tilde{\mu})$. Then at $\tilde{\mu}$, $0 < dQ^1/d\mu < dQ^0/d\mu$. This is a contradiction because $\lim_{\mu \rightarrow 0} Q^1(\mu) = \lim_{\mu \rightarrow 0} Q^0(\mu)$.

OA.2.2 Proof of Lemma 11

Let $h^\theta(\mu) = f^\theta(\mu)/(1 - F^\theta(\mu))$ denote the hazard rate conditional on θ . The posterior distribution conditional on $\theta = 0$ satisfies the definition of the ULR order: $F^0(\mu) \succ_{\text{ULR}} \hat{F}^0(\mu)$. Then $h^0(\mu) > \hat{h}^0(\mu)$ for $\mu \geq 1/2$ (Hopkins and Kornienko, 2007, Corollary 1). The ULR order implies the ex ante distribution \hat{F} is a mean-preserving spread of F (Hopkins and Kornienko, 2007, Proposition 1), so $F^1(\mu) + F^0(\mu) > \hat{F}^1(\mu) + \hat{F}^0(\mu)$ for $\mu \geq 1/2$. It then follows from Lemma 10 that $F^1(\mu) > \hat{F}^1(\mu)$.

OA.2.3 Proof of Lemma 12

For any two distributions $F \succ_{\text{ULR}} \hat{F}$, f/\hat{f} is unimodal. The likelihood ratio of F and $(1 - \lambda)F + \lambda\hat{F}$ is $f/((1 - \lambda)f + \lambda\hat{f})$ and the likelihood ratio of $(1 - \lambda)F + \lambda\hat{F}$ and \hat{F} is $((1 - \lambda)f + \lambda\hat{f})/\hat{f}$. Both are unimodal as implied by that f/\hat{f} is unimodal.

$F \succ_{\text{ULR}} \hat{F}$ implies the mean of F is (weakly) higher than the mean of \hat{F} . So the mean of F is (weakly) higher than the mean of $(1 - \lambda)F + \lambda\hat{F}$, which is (weakly) greater than the mean of \hat{F} . The result follows.

OA.2.4 Proof of Claim 4

The proof is mostly algebraic. For conciseness, I omit the argument of the functions. After some rearranging, \mathcal{V} can be written in terms of h ,

$$\mathcal{V} = \underbrace{q \left(1 - \frac{1-\mu}{\mu}\right)}_{=:b} \underbrace{- q \left(1 - \frac{1-\mu}{\mu}\right) \left(\frac{1-\mu}{\mu} \frac{1-F^1}{F^1}\right)}_{=:a} h.$$

That is, $\mathcal{V} = ah + b$. Let the superscript denote the (partial) derivative. Then $h^\lambda/h^\mu - \mathcal{V}^\lambda/\mathcal{V}^\mu = (h^\lambda/h^\mu)(a^\mu h + b^\mu)/\mathcal{V}^\mu - (a^\lambda h + b^\lambda)/\mathcal{V}^\mu$. Because $\mathcal{V}^\mu > 0$, $a^\mu h + b^\mu > -ah^\mu > 0$, showing Claim 4 is equivalent to showing $h^\lambda/h^\mu < (a^\lambda h + b^\lambda)/(a^\mu h + b^\mu)$. I prove the following chain of inequality: for all $\mu \geq 1/2$, $h^\lambda/h^\mu < q^\lambda/q^\mu < (a^\lambda h + b^\lambda)/(a^\mu h + b^\mu)$.

For the first inequality $h^\lambda/h^\mu < q^\lambda/q^\mu$, let $q = 1/(1 + m + dh)$ where

$$q = 1 / \left(1 + \underbrace{\frac{1-\mu}{\mu} \frac{1}{F^1}}_{=:m} - \underbrace{\left(\frac{1-\mu}{\mu}\right)^2 \frac{1-F^1}{F^1} h}_{=:d} \right).$$

It reduces to showing $h^\lambda/h^\mu - q^\lambda/q^\mu = (h^\lambda/h^\mu)(1 - h^\mu d/q^\mu) - (m^\lambda + d^\lambda h)/q^\mu < 0$. $h^\lambda < 0$ (Lemma 10), $h^\mu > 0$, $q^\mu > 0$, and $d < 0$, so $(h^\lambda/h^\mu)(1 - h^\mu d/q^\mu) < 0$. Note that $d = -m(1 - \mu)/\mu + ((1 - \mu)/\mu)^2$. Because $(1 - F^0)/(1 - F^1) < 1$ (MLRP) and $m^\lambda > 0$ (Lemma 11), $d^\lambda h = -m^\lambda(1 - F^0)/(1 - F^1) > -m^\lambda$, so $(m^\lambda + d^\lambda h)/q^\mu > 0$.

For the second inequality $q^\lambda/q^\mu < (a^\lambda h + b^\lambda)/(a^\mu h + b^\mu)$, the right-hand side is

$$\frac{\overbrace{q^\lambda \left(2 - \frac{1}{\mu}\right) \left(1 - \frac{1-F^0}{F^1}\right)}^{=: \alpha} - \overbrace{\left(\frac{1-F^1}{F^1}\right)^\lambda \frac{1-\mu}{\mu} b h}^{=: \beta}}{\underbrace{q^\mu \left(2 - \frac{1}{\mu}\right) \left(1 - \frac{1-F^0}{F^1}\right)}_{=: \alpha} + \underbrace{\left(2 - \frac{1}{\mu}\right)^\mu q \left(1 - \frac{1-F^0}{F^1}\right) - \left(\frac{1-\mu}{\mu} \frac{1-F^1}{F^1}\right)^\mu b h}_{=: \eta}}.$$

It reduces to showing $q^\lambda/q^\mu - (a^\lambda h + b^\lambda)/(a^\mu h + b^\mu) = (q^\lambda/q^\mu)\eta/(q^\mu\alpha + \eta) - \beta/(q^\mu\alpha + \eta) < 0$. Because $q^\mu\alpha + \eta > 0$, it is equivalent to $q^\mu/q^\lambda - \eta/\beta > 0$. Writing out all the terms, this inequality follows from Lemma 10, Lemma 11, MLRP, IHRP, and symmetry.

OA.3 Omitted Proofs for **Section 5**

OA.3.1 Proof of **Theorem 2**

Equilibrium conditions

Leader-follower continuation game. Introducing a flow cost for the leader does not affect the follower's incentive. Same as the no-flow-cost case, the follower's first-order condition implies $x'(t) = \phi(x(t), y(t))$, where

$$\phi(x, y) := -r \left(\frac{\rho_0 f^1(y)(1 - F^1(x))(H - c) - (1 - \rho_0)f^0(y)(1 - F^0(x))(L + c)}{\rho_0 f^1(y)f^1(x)(H - c) - (1 - \rho_0)f^0(y)f^0(x)(L + c)} \right).$$

For leader of type x , same as before, denote his belief at the beginning of the leader-follower continuation game by $q_L(x) = \Pr(\theta = 1 | x, s_F < y(0))$. His expected payoff from disinvesting at t is

$$\begin{aligned} \mathcal{L}(x, t) = & q_L(x) \\ & \cdot \left(\int_0^t -y'(\tau) \frac{f^1(y(\tau))}{F^1(y(0))} \left(e^{-r\tau} H - \int_0^\tau e^{-r\tilde{\tau}} \eta d\tilde{\tau} \right) d\tau - \frac{F^1(y(t))}{F^1(y(0))} \int_0^t e^{-r\tilde{\tau}} \eta d\tilde{\tau} \right) \\ & - (1 - q_L(x)) \\ & \cdot \left(\int_0^t -y'(\tau) \frac{f^0(y(\tau))}{F^0(y(0))} \left(e^{-r\tau} L + \int_0^\tau e^{-r\tilde{\tau}} \eta d\tilde{\tau} \right) d\tau + \frac{F^0(y(t))}{F^0(y(0))} \int_0^t e^{-r\tilde{\tau}} \eta d\tilde{\tau} \right). \end{aligned}$$

The first-order condition implies $y'(t) = \psi(x(t), y(t))$, where

$$\psi(x, y) := -\eta \left(\frac{\rho_0 f^1(x)F^1(y) + (1 - \rho_0)f^0(x)F^0(y)}{\rho_0 f^1(x)f^1(y)H - (1 - \rho_0)f^0(x)f^0(y)L} \right).$$

Initial conditions. With strictly monotonic strategies, the flow cost does not affect the initial conditions. So the same as the no-flow cost case, $y(0) < z = x(0)$ and z 's indifference condition implies $W_0(x(0), y(0)) = c$, where

$$W_0(x, y) := \frac{\rho_0 f^1(x)(F^1(x) - F^1(y))H}{\rho_0 f^1(x)F^1(x) + (1 - \rho_0)f^0(x)F^0(x)} - \frac{(1 - \rho_0)f^0(x)(F^0(x) - F^0(y))L}{\rho_0 f^1(x)F^1(x) + (1 - \rho_0)f^0(x)F^0(x)}.$$

Optimality

To show optimality, one needs to show (i) $\mathcal{F}(y, t)$ is single-peaked in t , (ii) $\mathcal{L}(x, t)$ is single-peaked in t , and (iii) all types above z invest and all types below do not. (i) is

the same as the no-flow-cost case. The following lemma establishes (ii) holds. Given (i) and (ii), the proof of (iii) is the same as the no-flow-cost case.

Lemma OA.1. *For a fixed x , $\mathcal{L}(x, t)$ is single-peaked in t .*

Proof. The proof is analogous to the proof of [Lemma 7](#). To simplify notation, define

$$M(x, t) := \frac{q_L(x)}{F^1(y(0))}(-y'(t))f^1(y(t))H - \frac{1 - q_L(x)}{F^0(y(0))}(-y'(t))f^0(y(t))L,$$

$$N(x, t) := \left(\frac{q_L(x)}{F^1(y(0))}F^1(y(t)) + \frac{1 - q_L(x)}{F^0(y(0))}F^0(y(t)) \right) \eta.$$

In words, $e^{-rt}M(x, t)dt$ is type x 's marginal benefit from waiting for dt before disinvesting and $e^{-rt}N(x, t)dt$ is the marginal cost. Let the subscript i denote the partial derivative with respect to the i -th argument. The first-order condition of \mathcal{L} implies $M(x(t), t) = N(x(t), t)$. Because strategies are strictly monotone and everywhere differentiable, at each t , there exists one and only one type whose first-order condition is satisfied at t . Denote the type whose first-order condition is satisfied at t^* by x^* , that is, $M(x^*, t^*) = N(x^*, t^*)$. Suppose x^* mimics the behavior of type \hat{x} by stopping at \hat{t} . Because $M(x, t)$ is differentiable in x , by the fundamental theorem of calculus,

$$M(x^*, \hat{t}) = M(\hat{x}, \hat{t}) + \int_{\hat{x}}^{x^*} M_1(x, \hat{t})dx = N(\hat{x}, \hat{t}) + \int_{\hat{x}}^{x^*} M_1(x, \hat{t})dx,$$

where $M_1(x, \hat{t}) = dM(x, \hat{t})/dx$. The second equality follows from \hat{x} 's first-order condition $M(\hat{x}, \hat{t}) = N(\hat{x}, \hat{t})$. By MLRP, $q_L(x)$ is decreasing in x and because $y'(t) < 0$, so $M_1(x, \hat{t}) > 0$. Thus, if $\hat{x} < x^*$, then

$$M(x^*, \hat{t}) = N(\hat{x}, \hat{t}) + \int_{\hat{x}}^{x^*} M_1(x, \hat{t})dx > N(\hat{x}, \hat{t}) > N(x^*, \hat{t}),$$

where the first inequality follows from $\int_{\hat{x}}^{x^*} M_1(x, \hat{t})dx > 0$, and the second inequality follows from that N is decreasing in x because of MLRP and $y(t) < y(0)$. Similarly, if $\hat{x} > x^*$, then $\int_{\hat{x}}^{x^*} M_1(x, \hat{t})dx < 0$, so

$$M(x^*, \hat{t}) = N(\hat{x}, \hat{t}) + \int_{\hat{x}}^{x^*} M_1(x, \hat{t})dx < N(\hat{x}, \hat{t}) < N(x^*, \hat{t}).$$

$x(t)$ is increasing, so $\hat{x} < (>)x^*$ is equivalent to $\hat{t} < (>)t^*$. The above argument shows

$M(x^*, \hat{t}) - N(x^*, \hat{t}) > 0$ for all $\hat{t} < t^*$ and $M(x^*, \hat{t}) - N(x^*, \hat{t}) < 0$ for all $\hat{t} > t^*$. \square

Existence

In any dynamic equilibrium in strictly monotonic and differentiable strategies,

- (i) by optimality, players must get strictly positive payoff;
- (ii) strategies are strictly monotone: $x'(t) > 0$ and $y'(t) < 0$ for all $t \geq 0$;
- (iii) strategies are differentiable for all $t \geq 0$ and $x(t), y(t) \in (0, 1)$.

(i) In the leader-follower game, for the leader, disinvesting at $t = 0$ generates payoff 0 for any types of the leader, that is, $\mathcal{L}(x, 0) = 0$ for all $x \geq x(0)$. By [Lemma OA.1](#), $\mathcal{L}(x, t)$ is single-peaked in t , so by optimality, if a type optimally disinvests at $t > 0$, he must expect to get a strictly higher payoff than disinvesting at $t = 0$. That is, $\mathcal{L}(x(t), t) > \mathcal{L}(x(t), 0) = 0$ for all $x(t) > x(0)$. For the follower, $\mathcal{F}(y(t), t) > 0$ if and only if

$$\frac{\rho_0}{1 - \rho_0} \frac{f^1(y(t))}{f^0(y(t))} \frac{1 - F^1(x(t))}{1 - F^0(x(t))} > \frac{L + c}{H - c}. \quad (\text{OA.2})$$

I now show players' expected payoff at the beginning of the game is positive. Note that

$$\frac{\rho_0}{1 - \rho_0} \frac{f^1(x(0))}{f^0(x(0))} \frac{1 - F^1(x(0))}{1 - F^0(x(0))} > \frac{\rho_0}{1 - \rho_0} \frac{f^1(y(0))}{f^0(y(0))} \frac{1 - F^1(x(0))}{1 - F^0(x(0))} > \frac{L + c}{H - c},$$

where the first inequality follows from $x(0) > y(0)$, and the second inequality follows from evaluating [\(OA.2\)](#) at $t = 0$. This implies z 's ex ante expected payoff is strictly positive. By MLRP, all types above z receive strictly positive payoffs. Types below z do not invest at the beginning of the game so their payoff is at least 0.

(ii) $y'(t) < 0$ if and only if

$$\frac{\rho_0}{1 - \rho_0} \frac{f^1(y(t))}{f^0(y(t))} \frac{f^1(x(t))}{f^0(x(t))} > \frac{L}{H}. \quad (\text{OA.3})$$

Given [\(OA.2\)](#), $x'(t) > 0$ if and only if

$$\frac{\rho_0}{1 - \rho_0} \frac{f^1(y(t))}{f^0(y(t))} \frac{f^1(x(t))}{f^0(x(t))} < \frac{L + c}{H - c}. \quad (\text{OA.4})$$

(iii) Because $\phi(\cdot, \cdot)$ and $\psi(\cdot, \cdot)$ are autonomous first-order differential equations and are continuous for all (x, y) such that $\phi(x, y) > 0$ and $\psi(x, y) < 0$, and $x(t)$ and $y(t)$ are bounded, so as $t \rightarrow \infty$, $x'(t) \rightarrow 0$ and $y'(t) \rightarrow 0$. Note that $x'(t) = 0$ and $y'(t) = 0$

if and only if $x(t) = 1$ and $y(t) = 0$. So $\phi(x(t), y(t)) \rightarrow 0$ and $\psi(x(t), y(t)) \rightarrow 0$ if and only if $x(t) \rightarrow 1$ and $y(t) \rightarrow 0$.

Define $\mathcal{D} \subset (0, 1)^2$ and $\mathcal{D}_0 \subset (0, 1)^2$ as

$$\mathcal{D} := \{(x, y) : (\text{OA.2}), (\text{OA.3}) \text{ and } (\text{OA.4}) \text{ hold}\},$$

$$\mathcal{D}_0 := \mathcal{D} \cap \{(x, y) : x > y \text{ and } V(x, y) = c\}.$$

In words, if a solution $(x(t), y(t))$ to the differential system (8) is an equilibrium, then it must be that $(x(t), y(t)) \in \mathcal{D}$ for all $t \geq 0$ with initial values $(x(0), y(0)) \in \mathcal{D}_0$.

It is helpful to consider the (x, y) -plane and the differential equation

$$y'(x) = \Upsilon(x, y) := \frac{\psi(x, y)}{\phi(x, y)}, \quad \forall (x, y) \in \mathcal{D}. \quad (\text{OA.5})$$

By definition, $\Upsilon(x, y)$ is continuous in (x, y) for all $(x, y) \in \mathcal{D}$. An equilibrium is a solution $y(x)$ to the differential equation (OA.5) in \mathcal{D} with $y(x) < x$ that goes through a point in \mathcal{D}_0 and converges to 0 as x goes to 1. Showing an equilibrium exists and is unique is equivalent to showing such solution exists and is unique. In what follows, **Lemma OA.2** shows there exists a trajectory in \mathcal{D} that converges to 0 as x goes to 1. Under parametric restriction (OA.12), this trajectory is unique. **Lemma OA.3** shows this (unique) trajectory goes through one and only one point in \mathcal{D}_0 for $y(x) < x$. Thus the equilibrium is unique.

Figure OA.1 illustrates the unique equilibrium trajectory (red arrowed curve) which goes through exactly one point in \mathcal{D}_0 and converges to the point $(1, 0)$. All other trajectories (black arrowed curves) will diverge to the boundaries of \mathcal{D} . **Figure OA.1** also displays annotations that facilitate the rest of the proof.

Lemma OA.2. *For any feasible parameters, there exists a solution $y(x)$ to the differential equation (OA.5) in \mathcal{D} with $y(x) \rightarrow 0$ as $x \rightarrow 1$.*

Proof. Consider the boundaries of \mathcal{D} . For any fixed $x \in (0, 1)$, let $\beta_F(x)$ be such that

$$\frac{\rho_0}{1 - \rho_0} \frac{f^1(\beta_F(x))}{f^0(\beta_F(x))} \frac{1 - F^1(x)}{1 - F^0(x)} = \frac{L + c}{H - c}, \quad (\text{OA.6})$$

$\beta_f(x)$ be such that

$$\frac{\rho_0}{1 - \rho_0} \frac{f^1(\beta_f(x))}{f^0(\beta_f(x))} \frac{f^1(x)}{f^0(x)} = \frac{L}{H}, \quad (\text{OA.7})$$

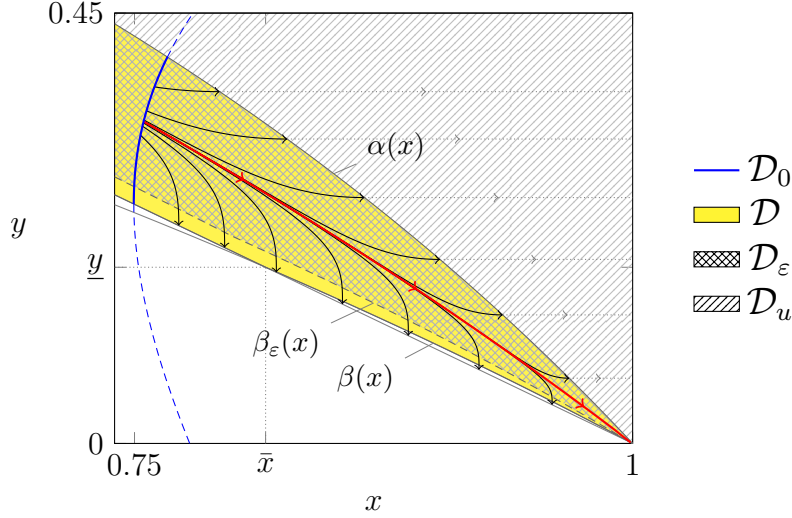


Figure OA.1: Equilibrium trajectory (red arrowed curve) and sample trajectories (non-equilibrium, black arrowed curves) to the differential system (8) for $\rho_0 = 1/2, H = L = 1, r = 1/5, c = 0.38, \eta = 1/20$ and posterior beliefs distributed according to $Beta(1 + \theta, 1 + (1 - \theta))$.

and $\alpha(x)$ be such that

$$\frac{\rho_0}{1 - \rho_0} \frac{f^1(\alpha(x))}{f^0(\alpha(x))} \frac{f^1(x)}{f^0(x)} = \frac{L + c}{H - c}. \quad (\text{OA.8})$$

Finally, define

$$\beta(x) := \max_{x \in (0,1)} \{\beta_F(x), \beta_f(x)\}.$$

By IHRP, $\beta_f(x)$ and $\beta_F(x)$ intersect at most once for $x \in (0, 1)$.

Claim OA.1. (i) \mathcal{D} is non-empty. (ii) $(1, 0) \in \text{cl}(\mathcal{D})$ and $(0, 1) \in \text{cl}(\mathcal{D})$.

Proof. (i) Fix $x \in (0, 1)$. By MLRP, the left-hand side of (OA.6) evaluated at any $(x', y') > (x, \beta_F(x))$ is strictly higher than $(L + c)/(H - c)$, the left-hand side of (OA.7) evaluated at any $(x', y') > (x, \beta_f(x))$ is strictly higher than L/H , and the left-hand side of (OA.8) evaluated at any $(x', y') < (x, \alpha(x))$ is strictly lower than $(L + c)/(H - c)$. $\alpha(x) > \beta(x)$ for all $x \in (0, 1)$. So \mathcal{D} is non-empty.

(ii) Fix $x \in (0, 1)$. Consider (OA.6). Take the limit of both sides as $x \rightarrow 1$. The right-hand side is constant at $(L + c)/(H - c)$. On the left-hand side, because $\lim_{x \rightarrow 1} \frac{1 - F^1(x)}{1 - F^0(x)} = \lim_{x \rightarrow 1} \frac{f^1(x)}{f^0(x)} = \infty$, it must be $f^1(\beta_F(x))/f^0(\beta_F(x)) \rightarrow 0$, which means $\beta_F(x) \rightarrow 0$. The same argument applies for equations (OA.7) and (OA.8). This implies $(1, 0) \in \text{cl}(\mathcal{D})$. An analogous argument shows $(0, 1) \in \text{cl}(\mathcal{D})$. \square

By definition, for all $(x, y) \in \mathcal{D}$, $\psi(x, y) < 0$ and $\phi(x, y) > 0$, so $\Upsilon(x, y) < 0$. Define $\mathcal{D}_u \in (0, 1)^2$ (the subscript u stands for “upper”) as

$$\mathcal{D}_u := \left\{ (x, y) : \frac{\rho_0}{1 - \rho_0} \frac{f^1(y)}{f^0(y)} \frac{f^1(x)}{f^0(x)} \geq \frac{L + c}{H - c} \right\}.$$

In words, \mathcal{D}_u is the set of points in the (x, y) -plane that are equal to or above $\alpha(x)$. By definition and the continuity of the distribution functions, $\mathcal{D} \cup \mathcal{D}_u$ is connected. For any fixed $x \in (0, 1)$, as $y \rightarrow \alpha(x)$, $\Upsilon(x, y) \rightarrow 0$. Let $\Upsilon(x, y) = 0$ for all $(x, y) \in \mathcal{D}_u$. Then $\Upsilon(x, y)$ is continuous in (x, y) for all $(x, y) \in \mathcal{D} \cup \mathcal{D}_u$. Apply the implicit function theorem to (OA.8), MLRP implies for all feasible parameters and $x \in (0, 1)$,

$$\alpha'(x) < 0 = \Upsilon(x, \alpha(x)).$$

This means $\alpha(x)$ is a strong lower fence (or lower solution, see [Hubbard and West, 1991](#), Section 1.3, or [Teschl, 2012](#), Section 1.5) for the differential equation

$$y'(x) = \Upsilon(x, y) = \begin{cases} \psi(x, y)/\phi(x, y) & (x, y) \in \mathcal{D} \\ 0 & (x, y) \in \mathcal{D}_u \end{cases}. \quad (\text{OA.9})$$

Consider an ε -variation of $\beta_F(x)$ and $\beta_f(x)$. Let $\beta_{F,\varepsilon}(x)$ be such that

$$\frac{\rho_0}{1 - \rho_0} \frac{f^1(\beta_{F,\varepsilon}(x))}{f^0(\beta_{F,\varepsilon}(x))} \frac{1 - F^1(x)}{1 - F^0(x)} = \frac{L + c}{H - c} + \varepsilon, \quad (\text{OA.10})$$

and $\beta_{f,\varepsilon}(x)$ be such that

$$\frac{\rho_0}{1 - \rho_0} \frac{f^1(\beta_{f,\varepsilon}(x))}{f^0(\beta_{f,\varepsilon}(x))} \frac{f^1(x)}{f^0(x)} = \frac{L}{H} + \varepsilon. \quad (\text{OA.11})$$

Define

$$\beta_\varepsilon(x) := \max_{x \in (0,1)} \{\beta_{F,\varepsilon}(x), \beta_{f,\varepsilon}(x)\},$$

$$\mathcal{D}_\varepsilon := \{(x, y) : x \in (0, 1) \text{ and } \beta_\varepsilon(x) \leq y < \alpha(x)\}.$$

By MLRP, for all $x \in (0, 1)$, $\beta_{F,\varepsilon}(x) < \alpha(x)$. For all $\varepsilon < (L + c)/(H - c) - L/H$, $\beta_{f,\varepsilon}(x) < \alpha(x)$. By the same argument as [Claim OA.1](#), \mathcal{D}_ε is non-empty, and the points $(1, 0)$ and $(0, 1)$ are in the closure of \mathcal{D}_ε . Moreover, $\mathcal{D}_\varepsilon \cup \mathcal{D}_u$ is connected and $\Upsilon(x, y)$ is continuous in (x, y) for all $(x, y) \in \mathcal{D}_\varepsilon \cup \mathcal{D}_u$.

Apply the implicit function theorem to (OA.10) and (OA.11), MLRP implies that for all feasible parameters and any $\varepsilon > 0$, $\beta'_{F,\varepsilon}(x)$ and $\beta'_{f,\varepsilon}(x)$ are both finite and negative. Therefore $\beta'_\varepsilon(x) > -\infty$ for all $x \in (0, 1)$.

Claim OA.2. There exists $\hat{\varepsilon} > 0$ such that $\Upsilon(x, \beta_\varepsilon(x)) < \beta'_\varepsilon(x)$ for all x .

Proof. For all $x \in (0, 1)$, by definition, as $\varepsilon \rightarrow 0$, $\beta_\varepsilon(x) \rightarrow \beta(x)$, which implies $\Upsilon(x, \beta_\varepsilon(x)) \rightarrow -\infty$. So for any x , there exists $\varepsilon(x) > 0$ (ε might depend on x) such that for all $\varepsilon < \varepsilon(x)$, $\Upsilon(x, \beta_\varepsilon(x)) < \beta'_\varepsilon(x)$. Let $\hat{\varepsilon} := \inf_{x \in (0, 1)} \varepsilon(x)$. It remains to show $\hat{\varepsilon} > 0$. Suppose $\hat{\varepsilon} = 0$. Then there exists a sequence ε_n with $\varepsilon_n \rightarrow 0$ such that for each ε_n there exists x_n such that $\Upsilon(x_n, \beta_{\varepsilon_n}(x_n)) \geq \beta'_{\varepsilon_n}(x_n)$. This is a contradiction because for all x_n , $\beta'_{\varepsilon_n}(x_n) > -\infty$ but as $\varepsilon_n \rightarrow 0$, $\Upsilon(x_n, \beta_{\varepsilon_n}(x_n)) \rightarrow -\infty$. \square

This means $\beta_\varepsilon(x)$ is a strong upper fence (or upper solution) for the differential equation (OA.9). Therefore, in $\mathcal{D}_\varepsilon \cup \mathcal{D}_u$, there exists a solution $y(x)$ to the differential equation (OA.5) with $\beta_\varepsilon(x) \leq y(x) \leq \alpha(x)$ for all $x \in (0, 1)$ (see Hubbard and West, 1991, Theorem 1.4.4, or Teschl, 2012, Lemma 1.2).

The above argument establishes there exists a solution in $\mathcal{D}_\varepsilon \cup \mathcal{D}_u$. It remains to show that the solution is within \mathcal{D}_ε (and thus within \mathcal{D}), not in \mathcal{D}_u . This boils down to showing that solutions in \mathcal{D}_u do not converge to 0 as $x \rightarrow 1$. This follows from the definition that $y'(x) = 0$ for all $(x, y) \in \mathcal{D}_u$. So for any $(x, y(x)) \in \mathcal{D}_u$ that solves the differential equation (OA.9), $y(x) > 0$ for all x . \square

Uniqueness

Assumption. Assume the following condition holds:

$$\forall (x, y) \in \mathcal{D}, \quad \partial \Upsilon(x, y) / \partial y \geq 0. \quad (\text{OA.12})$$

The uniqueness of a global condition can be established if the primitives satisfy the above condition. It can be numerically verified that (OA.12) is satisfied if f^θ is induced by signals distributed according to the Beta distributions or the Normal distributions. Moreover, by definition, as $x \rightarrow 1$, $\alpha(x) \rightarrow 0$ and $\beta_\varepsilon(x) \rightarrow 0$, so

$$\lim_{x \rightarrow 1} |\alpha(x) - \beta_\varepsilon(x)| = 0. \quad (\text{OA.13})$$

Conditions (OA.12) and (OA.13) imply the solution is unique in \mathcal{D}_ε (see Hubbard and West, 1991, Theorem 1.4.5, or Teschl, 2012, Section 1.5).

The above argument establishes the unique solution is in \mathcal{D}_ε . It remains to show this solution is unique in \mathcal{D} . Because $\mathcal{D} = \mathcal{D}_\varepsilon \cup \{(x, y) : x \in (0, 1) \text{ and } \beta(x) < y < \beta_\varepsilon(x)\}$, it boils down to showing there does not exist a solution in the set $\{(x, y) : x \in (0, 1) \text{ and } \beta(x) < y < \beta_\varepsilon(x)\}$. For all $y(x)$ such that $\beta(x) < y(x) < \beta_\varepsilon(x)$, $y'(x) \rightarrow -\infty$, which implies for all $x \in (0, 1)$, $y(x) \rightarrow \beta(x) > 0$.

Denote this unique solution by $\hat{y}(x)$. I prove there exists a unique set of initial values satisfying $\hat{y}(x)$. This is summarized in the following lemma.

Lemma OA.3. *There exists a unique $(x_0, y_0) \in \mathcal{D}_0$ such that $y_0 = \hat{y}(x_0)$.*

Proof. To simplify notation, define

$$\ell(x, y) := \frac{\rho_0}{1 - \rho_0} \frac{f^1(y)}{f^0(y)} \frac{f^1(x)}{f^0(x)}.$$

Recall that \mathcal{D}_0 is the set of points $(x, y) \in \mathcal{D}$ that satisfies the equation $W_0(x, y) = c$. Solve $W_0(x, y) = c$ for y in terms of x and denote the solution by $y_{W_0}(x)$. By [Claim 6](#) (iii) and (iv), $y_{W_0}(x)$ is increasing and continuous in x for all x such that $y_{W_0}(x) < x$.

By a change of variable, [Lemma OA.2](#) shows $\hat{y}(x)$ also converges to 1 as $x \rightarrow 0$. So $\hat{y}(x)$ is a strictly decreasing function that converges to 1 as $x \rightarrow 0$ and converges to 0 as $x \rightarrow 1$, and satisfies $\ell(x, \hat{y}(x)) \in (L/H, (L + c)/(H - c))$ for all $x \in (0, 1)$. So points in \mathcal{D}_0 constitute a strictly increasing and continuous function that starts at a point below $\hat{y}(x)$, and ends at a point above $\hat{y}(x)$. The result follows. \square

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