

Appendix to: “Conversations Among Competitors”

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The proof of Proposition 4 builds on that of Proposition 5, so these two results are proved in reverse order.

A. PROOF OF PROPOSITION 5:

Consider a given path of play in the equilibrium described in Proposition 3. Until there are more than k^* signals available for decoding, a new conversation is initiated each time a conversant fails to generate an additional signal. Since player A is assumed to start with an initial signal, c is the random number of failures in a series of independent Bernoulli (p) trials before there are $I(k^*)$ successes. A random variable such as c with

$$(A.1) \quad \Pr(c = i) = \binom{i + I(k^*) - 1}{I(k^*) - 1} p^{I(k^*)} (1 - p)^i \text{ for } i = 0, 1, 2, \dots$$

is said to follow the negative binomial distribution. It is then straightforward to show that $E(c) = I(k^*)(1 - p)/p$ (see George Casella and Roger L. Berger (2002), p. 95-96).

Next, note that the random variable k follows a geometric distribution with

$$(A.2) \quad \Pr(k = i) = (1 - p)p^{i-1} \text{ for } i = 1, 2, \dots$$

A simple calculation confirms that $E(k) = 1/(1 - p)$ (see Casella and Berger (2002), p. 97).

B. PROOF OF PROPOSITION 4:

First, recall that $\text{cov}(c, k) = E(ck) - E(c)E(k)$. By Proposition 5, we have $E(c)E(k) = I(k^*)/p$. To calculate $E(ck)$, we can use the fact that $E(ck) = E[E(c/k)k]$ where $E(c/k)$ denotes the conditional expectation of c given k . Next note that:

$$(A.3) \quad E(c | k) = \begin{cases} 1 + (I(k^*) + 1 - k) \frac{(1-p)}{p} & k \leq I(k^*) \\ 0 & k \geq I(k^*) + 1. \end{cases}$$

Thus, for $k \leq I(k^*)$, we carry out Bernoulli (p) trials until there are $I(k^*) + 1 - k$ successes, so the expression for $E(c/k)$ follows from Proposition 5 (we must add one since we are effectively conditioning on $c \geq 1$); while for $k \geq I(k^*) + 1$, B will not initiate a conversation with C so that $E(c/k) = 0$. Note that $E(c/k)$ is decreasing in k . Calculating, we then obtain:

$$(A.4) \quad E[E(c | k)k] = \sum_{i=1}^{I(k^*)} \left[\left(1 + (I(k^*) + 1 - i) \frac{(1-p)}{p} \right) i \right] (1-p)p^{i-1} = \frac{I(k^*)}{p} - \frac{1-p^{I(k^*)}}{1-p},$$

so that $\text{cov}(c, k) = -(1 - p^{I(k^*)}) / (1 - p) < 0$. After computing $\text{var}(c)$ and $\text{var}(k)$ (see Casella and Berger (2002), p. 95-97), and rearranging the resulting expression for

$\rho_{ck} = \text{cov}(c, k) / \sqrt{\text{var}(c) \text{var}(k)}$, we have:

$$(A.5) \quad \rho_{ck} = -\left(1 - p^{I(k^*)}\right) \sqrt{\frac{p}{I(k^*)(1-p)}} < 0.$$

C. COMPARATIVE STATICS PROPERTIES OF $E(c)$ and $\Phi\{E(c)\}$:

Let us begin with the comparative statics of k^* with respect to p and θ . Let

$$(A.6) \quad F(k, p, \beta, \theta) = \left[\frac{\theta(1-p) + p(1-\beta)}{1-p\beta} \right] \beta^k - \theta$$

and note that the function $k^*(p, \beta, \theta)$ is defined implicitly by $F(k^*(p, \beta, \theta), p, \beta, \theta) = 0$.

Since $\partial F(k, p, \beta, \theta) / \partial k < 0$, applying the Implicit Function Theorem and noting that

$$(A.7) \quad \frac{\partial F(k^*, p, \beta, \theta)}{\partial p} = \frac{(1-\theta)(1-\beta)}{(1-p\beta)^2} \beta^{k^*} > 0, \text{ and}$$

$$(A.8) \quad \frac{\partial F(k^*, p, \beta, \theta)}{\partial \theta} = \frac{(1-p)}{(1-p\beta)} \beta^{k^*} - 1 < 0,$$

we see that $\partial k^* / \partial p > 0$ and $\partial k^* / \partial \theta < 0$. Since $I(k^*)$ is weakly increasing in k^* , it follows immediately from the above results that $E(c) = I(k^*)(1-p)/p$ is weakly decreasing in θ .

Finally, recall that the upper bound on $E[c]$ is given by $\Phi\{E[c]\} = k^*(1-p)/p$.

Differentiating with respect to p yields

$$(A.9) \quad \frac{\partial \Phi\{E[c]\}}{\partial p} = \frac{\partial k^*}{\partial p} \frac{(1-p)}{p} - \frac{k^*}{p^2}.$$

Computing $\partial k^* / \partial p$ and noting that $(1-p)/(1-p\beta) < 1$, we have

$$(A.10) \quad \begin{aligned} \frac{\partial k^*}{\partial p} \frac{(1-p)}{p} &= \left[-\frac{1}{\ln[\beta]} \frac{(1-\theta)(1-\beta)}{(1-p\beta)[\theta(1-p) + p(1-\beta)]} \right] \frac{(1-p)}{p} \\ &< -\frac{1}{p \ln[\beta]} \left[\frac{(1-\theta)(1-\beta)}{\theta(1-p) + p(1-\beta)} \right]. \end{aligned}$$

Finally since $-\ln[x] > 1-x$ for $0 < x < 1$, it follows that

$$(A.11) \quad \begin{aligned} \frac{k^*}{p^2} &= \frac{1}{p^2} \frac{\log \left[\frac{(1-p\beta)\theta}{\theta(1-p) + p(1-\beta)} \right]}{\ln[\beta]} \\ &> -\frac{1}{p^2 \ln[\beta]} \left[1 - \frac{(1-p\beta)\theta}{\theta(1-p) + p(1-\beta)} \right] \\ &= -\frac{1}{p \ln[\beta]} \left[\frac{(1-\theta)(1-\beta)}{\theta(1-p) + p(1-\beta)} \right]. \end{aligned}$$

Combining these inequalities, it follows that $\partial \Phi\{E[c]\} / \partial p < 0$.

Reference

Casella, George, and Roger L. Berger. 2002. *Statistical Inference*. Duxbury Press.