

Online Supplement to

“A Social Interaction Model with Ordered Choices”

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A Proofs

Proof of Proposition 1. Under Assumption 1 (i), $\vec{h}(\cdot)$ is continuous, and thus the existence of a solution to the system of equations

$$\mathbf{y}^E = \vec{h}(\mathbf{y}^E), \tag{A.1}$$

is guaranteed by the Brouwer fixed-point theorem. By the contraction mapping theorem, the system of equations (A.1) has a unique solution if $\vec{h}(\cdot)$ is a contraction mapping with respect to some matrix norm $\|\cdot\|$. Thus, to show the desired result, we only need to show that, under Assumption 1 (ii), $\vec{h}(\cdot)$ is a contraction mapping.

First, we show that $\vec{h}(\cdot)$ is a contraction mapping with respect to a (submultiplicative) matrix norm $\|\cdot\|$, if $\|\vec{h}'(\cdot)\| < 1$ where $\vec{h}'(\mathbf{x}) = \partial\vec{h}(\mathbf{x})/\partial\mathbf{x}'$. Let $\vec{k}(t) = \vec{h}(\mathbf{x}_1 + t(\mathbf{x}_2 - \mathbf{x}_1))$. If $\|\vec{h}'(\cdot)\| < 1$, then

$$\begin{aligned} \|\vec{h}(\mathbf{x}_2) - \vec{h}(\mathbf{x}_1)\| &= \|\vec{k}(1) - \vec{k}(0)\| = \left\| \int_0^1 \vec{k}'(t) dt \right\| = \left\| \int_0^1 (\mathbf{x}_2 - \mathbf{x}_1) \vec{h}'(\mathbf{x}_1 + t(\mathbf{x}_2 - \mathbf{x}_1)) dt \right\| \\ &\leq \int_0^1 \|\mathbf{x}_2 - \mathbf{x}_1\| \cdot \|\vec{h}'(\mathbf{x}_1 + t(\mathbf{x}_2 - \mathbf{x}_1))\| dt < \|\mathbf{x}_2 - \mathbf{x}_1\|, \end{aligned}$$

i.e. $\vec{h}(\cdot)$ is a contraction mapping.

Next, as the (i, j) -th element of $\partial\vec{h}(\mathbf{y}^E)/\partial\mathbf{y}^{E'}$ is given by

$$\partial h_i(\mathbf{y}^E)/\partial y_j^E = \lambda w_{ij} \sum_{k=1}^{m-1} f(\alpha_{m-k} - \lambda \sum_{j=1}^n w_{ij} y_j^E - \mathbf{x}'_i \boldsymbol{\beta}),$$

we have

$$|\partial h_i(\mathbf{y}^E)/\partial y_j^E| \leq (m-1)|\lambda| \cdot |w_{ij}| \cdot \sup_u f(u).$$

Hence, under Assumption 1 (ii), either

$$\|\partial\vec{h}(\mathbf{y}^E)/\partial\mathbf{y}^{E'}\|_\infty \leq (m-1)|\lambda| \cdot \|\mathbf{W}\|_\infty \sup_u f(u) < 1,$$

or

$$\|\partial\vec{h}(\mathbf{y}^E)/\partial\mathbf{y}^{E'}\|_1 \leq (m-1)|\lambda| \cdot \|\mathbf{W}\|_1 \sup_u f(u) < 1.$$

Therefore, under Assumption 1 (ii), the system of equations (A.1) is a contraction mapping with respect to the $\|\cdot\|_\infty$ or $\|\cdot\|_1$ norm, whichever is a smaller matrix norm of \mathbf{W} . \square

Proof of Proposition 2. Given the network topology \mathbf{W} , observational equivalence requires

$$\Pr(y_i \leq k | \mathbf{W}, \mathbf{X}) = F(\alpha_k - \lambda \mathbf{w}_i \mathbf{y}^E - \mathbf{x}'_i \boldsymbol{\beta}) = F(\tilde{\alpha}_k - \tilde{\lambda} \mathbf{w}_i \tilde{\mathbf{y}}^E - \mathbf{x}'_i \tilde{\boldsymbol{\beta}}),$$

for $k = 1, \dots, m-1$ and for any \mathbf{X} in its support, where, under Assumption 1, \mathbf{y}^E and $\tilde{\mathbf{y}}^E$ are uniquely determined by

$$y_i^E = m - \sum_{k=1}^{m-1} F(\alpha_k - \lambda \mathbf{w}_i \mathbf{y}^E - \mathbf{x}'_i \boldsymbol{\beta}) \tag{A.2}$$

$$\tilde{y}_i^E = m - \sum_{k=1}^{m-1} F(\tilde{\alpha}_k - \tilde{\lambda} \mathbf{w}_i \tilde{\mathbf{y}}^E - \mathbf{x}'_i \tilde{\boldsymbol{\beta}}) \tag{A.3}$$

for $i = 1, \dots, n$, respectively. The model is identified if observational equivalence of $\boldsymbol{\delta}$ and $\tilde{\boldsymbol{\delta}}$ implies $\boldsymbol{\delta} = \tilde{\boldsymbol{\delta}}$. That is,

$$F(\alpha_k - \lambda \mathbf{w}_i \mathbf{y}^E - \mathbf{x}'_i \boldsymbol{\beta}) = F(\tilde{\alpha}_k - \tilde{\lambda} \mathbf{w}_i \tilde{\mathbf{y}}^E - \mathbf{x}'_i \tilde{\boldsymbol{\beta}}), \tag{A.4}$$

for $k = 1, \dots, m - 1$, implies $\boldsymbol{\delta} = \tilde{\boldsymbol{\delta}}$. If (A.4) holds for $k = 1, \dots, m - 1$, then the right hand sides of (A.2) and (A.3) are identical, i.e., $y_i^E = \tilde{y}_i^E$, for $i = 1, \dots, n$. Therefore, (A.4) can be rewritten as

$$F(\alpha_k - \lambda \mathbf{w}_i \mathbf{y}^E - \mathbf{x}'_i \boldsymbol{\beta}) = F(\tilde{\alpha}_k - \tilde{\lambda} \mathbf{w}_i \mathbf{y}^E - \mathbf{x}'_i \tilde{\boldsymbol{\beta}}), \quad (\text{A.5})$$

for $k = 1, \dots, m - 1$. Under Assumption 2, (A.5) implies

$$(\alpha_k - \tilde{\alpha}_k) \boldsymbol{\iota}_n + (\lambda - \tilde{\lambda}) \mathbf{W} \mathbf{y}^E + \mathbf{X}(\boldsymbol{\beta} - \tilde{\boldsymbol{\beta}}) = 0,$$

which, in turn, implies $\alpha_k = \tilde{\alpha}_k$, $\lambda = \tilde{\lambda}$ and $\boldsymbol{\beta} = \tilde{\boldsymbol{\beta}}$, for $k = 1, \dots, m - 1$. Hence, observational equivalence of $\boldsymbol{\delta}$ and $\tilde{\boldsymbol{\delta}}$ implies $\boldsymbol{\delta} = \tilde{\boldsymbol{\delta}}$, i.e., the model is identified. \square