

## C Supplemental Web Appendix: Details of Calculations (Not for Publication)

### Calculation details for proof of Lemma 4

To derive (24), as given in the proof of Lemma 4, observe that tedious calculation gives

$$\begin{aligned}
 \nabla \varphi_{\mathbf{A}}(\mathbf{A}) &= \begin{pmatrix} \sum_{j \neq 1} \frac{\exp(W'_{1j}\beta) \exp(W'_{1j}\beta + A_1(\beta))}{[\exp(-A_j(\beta)) + \exp(W'_{1j}\beta + A_1(\beta))]^2} & -\frac{\exp(W'_{12}\beta) \exp(-A_2(\beta))}{[\exp(-A_2(\beta)) + \exp(W'_{12}\beta + A_1(\beta))]^2} \\ \sum_{j \neq 1} \frac{\exp(W'_{1j}\beta)}{\exp(-A_j(\beta)) + \exp(W'_{1j}\beta + A_1(\beta))} & \sum_{j \neq 1} \frac{\exp(W'_{1j}\beta)}{\exp(-A_j(\beta)) + \exp(W'_{1j}\beta + A_1(\beta))} \\ -\frac{\exp(W'_{12}\beta) \exp(-A_1(\beta))}{[\exp(-A_1(\beta)) + \exp(W'_{12}\beta + A_2(\beta))]^2} & \sum_{j \neq 2} \frac{\exp(W'_{2j}\beta) \exp(W'_{2j}\beta + A_2(\beta))}{[\exp(A_j(\beta)) + \exp(W'_{2j}\beta + A_2(\beta))]^2} \\ \sum_{j \neq 2} \frac{\exp(W'_{2j}\beta)}{\exp(-A_j(\beta)) + \exp(W'_{2j}\beta + A_2(\beta))} & \sum_{j \neq 2} \frac{\exp(W'_{2j}\beta)}{\exp(-A_j(\beta)) + \exp(W'_{2j}\beta + A_2(\beta))} \\ \vdots & \vdots \\ -\frac{\exp(W'_{1N}\beta) \exp(-A_1(\beta))}{[\exp(-A_1(\beta)) + \exp(W'_{1N}\beta + A_N(\beta))]^2} & \dots \\ \sum_{j \neq N} \frac{\exp(W'_{Nj}\beta)}{\exp(-A_j(\beta)) + \exp(W'_{Nj}\beta + A_N(\beta))} & \dots \\ \dots & \dots \\ \dots & -\frac{\exp(W'_{1N}\beta) \exp(-A_N(\beta))}{[\exp(-A_N(\beta)) + \exp(W'_{1N}\beta + A_1(\beta))]^2} \\ \sum_{j \neq 1} \frac{\exp(W'_{1j}\beta)}{\exp(-A_j(\beta)) + \exp(W'_{1j}\beta + A_1(\beta))} & \dots \\ \dots & -\frac{\exp(W'_{2N}\beta) \exp(-A_N(\beta))}{[\exp(-A_N(\beta)) + \exp(W'_{2N}\beta + A_2(\beta))]^2} \\ \sum_{j \neq 2} \frac{\exp(W'_{2j}\beta)}{\exp(-A_j(\beta)) + \exp(W'_{2j}\beta + A_2(\beta))} & \dots \\ \vdots & \vdots \\ \dots & \dots \\ \dots & \sum_{j \neq N} \frac{\exp(W'_{Nj}\beta) \exp(W'_{Nj}\beta + A_N(\beta))}{[\exp(A_j(\beta)) + \exp(W'_{Nj}\beta + A_N(\beta))]^2} \\ \sum_{j \neq N} \frac{\exp(W'_{Nj}\beta)}{\exp(-A_j(\beta)) + \exp(W'_{Nj}\beta + A_N(\beta))} & \dots \end{pmatrix} \\
 &= \begin{pmatrix} \frac{\sum_{j \neq 1} r_{1j} p_{1j}}{\sum_{j \neq 1} r_{1j}} & -\frac{r_{12}(1-p_{12})}{\sum_{j \neq 1} r_{1j}} & \dots & -\frac{r_{1N}(1-p_{1N})}{\sum_{j \neq 1} r_{1j}} \\ -\frac{r_{21}(1-p_{12})}{\sum_{j \neq 2} r_{2j}} & \frac{\sum_{j \neq 2} r_{2j} p_{2j}}{\sum_{j \neq 2} r_{2j}} & \dots & -\frac{r_{2N}(1-p_{2N})}{\sum_{j \neq 2} r_{2j}} \\ \vdots & \vdots & \ddots & \vdots \\ -\frac{r_{N1}(1-p_{1N})}{\sum_{j \neq N} r_{Nj}} & -\frac{r_{2N}(1-p_{2N})}{\sum_{j \neq N} r_{Nj}} & \dots & \frac{\sum_{j \neq N} r_{Nj} p_{Nj}}{\sum_{j \neq N} r_{Nj}} \end{pmatrix},
 \end{aligned}$$

where the second equality follows from the definition

$$r_{ij}(\beta, \mathbf{A}, W_{ij}) = \frac{\exp(W'_{ij}\beta)}{\exp(-A_j) + \exp(W'_{ij}\beta + A_i)} = \exp(A_i) p_{ij},$$

and the relationships

$$\frac{\frac{\exp(W'_{ij}\beta) \exp(-A_j(\beta))}{[\exp(-A_j(\beta)) + \exp(W'_{ij}\beta + A_i(\beta))]^2}}{\sum_{j \neq i} \frac{\exp(W'_{ij}\beta)}{\exp(-A_j(\beta)) + \exp(W'_{ij}\beta + A_i(\beta))}} = \frac{r_{ij} (1 - p_{ij})}{\sum_{j \neq i} r_{ij}} = \frac{p_{ij} (1 - p_{ij})}{\sum_{j \neq i} p_{ij}},$$

and

$$\frac{\sum_{j \neq i} \frac{\exp(W'_{ij}\beta) \exp(W'_{ij}\beta + A_i(\beta))}{[\exp(-A_j(\beta)) + \exp(W'_{ij}\beta + A_i(\beta))]^2}}{\sum_{j \neq i} \frac{\exp(W'_{ij}\beta)}{\exp(-A_j(\beta)) + \exp(W'_{ij}\beta + A_i(\beta))}} = \frac{\sum_{j \neq i} r_{ij} p_{ij}}{\sum_{j \neq i} r_{ij}} = \frac{\sum_{j \neq i} p_{ij}^2}{\sum_{j \neq i} p_{ij}}.$$

## Calculation details for proof of Lemma 6

To derive the bound for  $R_p$  appearing in the proof of Lemma 6 observe that

$$\frac{\partial}{\partial \mathbf{A} \partial \mathbf{A}'} s_{\mathbf{A}ij}^{(p)}(\beta_0, \mathbf{A}(\beta_0)) = -p_{ij} (1 - p_{ij}) (1 - 2p_{ij}) T_{ij} T'_{ij} T_{p,ij}$$

and hence that  $\sum_{i=1}^N \sum_{j < i} \frac{\partial}{\partial \mathbf{A} \partial \mathbf{A}'} s_{\mathbf{A}ij}^{(p)}(\beta_0, \mathbf{A}(\beta_0))$  equals

$$- \begin{pmatrix} \sum_{j \neq 1} p_{1j} (1 - p_{1j}) (1 - 2p_{1j}) T_{p,1j} & \cdots & p_{1N} (1 - p_{1N}) (1 - 2p_{1N}) T_{p,1N} \\ \vdots & \ddots & \vdots \\ p_{1N} (1 - p_{1N}) (1 - 2p_{1N}) T_{p,1N} & \cdots & \sum_{j \neq N} p_{Nj} (1 - p_{Nj}) (1 - 2p_{Nj}) T_{p,Nj} \end{pmatrix}.$$

So that

$$\iota'_N \left[ \sum_{i=1}^N \sum_{j < i} \frac{\partial}{\partial \mathbf{A} \partial \mathbf{A}'} s_{\mathbf{A}ij}^{(p)}(\beta_0, \mathbf{A}(\beta_0)) \right] \iota_N = 2 \sum_{i=1}^N \sum_{j \neq i} p_{ij} (1 - p_{ij}) (1 - 2p_{ij}) T_{p,ij}.$$

Finally observe that  $\sum_{i=1}^N \sum_{j \neq i} T_{p,ij} = 2(N - 1)$ .

## Calculation details for proof of Theorem 1

To derive (32) use iterated expectations to show that

$$\begin{aligned}
L(\beta) &= \mathbb{E} [ |S_{ij,kl}| \{ S_{ij,kl} W'_{ij,kl} \beta - \ln [1 + \exp (S_{ij,kl} W'_{ij,kl} \beta)] \} ] \\
&= \Pr (S_{ij,kl} \in \{-1, 1\}) \mathbb{E} [ S_{ij,kl} W'_{ij,kl} \beta - \ln [1 + \exp (S_{ij,kl} W'_{ij,kl} \beta)] | S_{ij,kl} \in \{-1, 1\} ] \\
&= \Pr (S_{ij,kl} \in \{-1, 1\}) \\
&\quad \times \mathbb{E} [ \mathbb{E} [ S_{ij,kl} W'_{ij,kl} \beta - \ln [1 + \exp (S_{ij,kl} W'_{ij,kl} \beta)] | X_i, X_j, X_k, X_l, S_{ij,kl} \in \{-1, 1\} ] \\
&\quad | S_{ij,kl} \in \{-1, 1\} ].
\end{aligned}$$

Evaluating the innermost expectation then yields

$$\begin{aligned}
&\mathbb{E} [ S_{ij,kl} W'_{ij,kl} \beta - \ln [1 + \exp (S_{ij,kl} W'_{ij,kl} \beta)] | X_i, X_j, X_k, X_l, S_{ij,kl} \in \{-1, 1\} ] \\
&= \{ W'_{ij,kl} \beta - \ln [1 + \exp (W'_{ij,kl} \beta)] \} q_{ij,kl} \\
&\quad + \{ -W'_{ij,kl} \beta - \ln [1 + \exp (-W'_{ij,kl} \beta)] \} [1 - q_{ij,kl}] \\
&= \ln \{ q_{ij,kl} (\beta) \} q_{ij,kl} + \ln \{ 1 - q_{ij,kl} (\beta) \} [1 - q_{ij,kl}] \\
&= - \left\{ q_{ij,kl} \ln \left( \frac{q_{ij,kl}}{q_{ij,kl} (\beta)} \right) + [1 - q_{ij,kl}] \ln \left( \frac{1 - q_{ij,kl}}{1 - q_{ij,kl} (\beta)} \right) \right\} \\
&\quad + q_{ij,kl} \ln (q_{ij,kl}) + [1 - q_{ij,kl}] \ln (1 - q_{ij,kl}) \\
&= -D_{KL} (q_{ij,kl} \| q_{ij,kl} (\beta)) - \mathbf{S} (q_{ij,kl}).
\end{aligned}$$

Fixing  $i$  and  $j$  and averaging with respect to independent random draws  $k$  and  $l$ , from the population of agents, yields

$$\begin{aligned}
\mathbb{E} [ S_{ij,kl} | i, j, \mathbf{X}, \mathbf{A} ] &= D_{ij} \Pr (D_{kl} = 1, D_{ik} = 0, D_{jl} = 0 | i, j, \mathbf{X}, \mathbf{A}) \\
&\quad - (1 - D_{ij}) \Pr (D_{kl} = 0, D_{ik} = 1, D_{jl} = 1 | i, j, \mathbf{X}, \mathbf{A}) \\
&= D_{ij} \mathbb{E} [ p_{kl} (1 - p_{ik}) (1 - p_{jl}) | i, j, \mathbf{X}, \mathbf{A} ] \\
&\quad - (1 - D_{ij}) \mathbb{E} [ (1 - p_{kl}) p_{ik} p_{jl} | i, j, \mathbf{X}, \mathbf{A} ].
\end{aligned} \tag{55}$$

An implication of (55) is that  $C(\bar{s}_{2,ij}, \bar{s}_{2,kl} | \mathbf{X}, \mathbf{A}) = 0$  unless  $ij$  and  $kl$  correspond to the same dyad. This is an implication of independent edge formation *conditional* on  $\mathbf{X}$  and  $\mathbf{A}$ .

To derive (35) observe that

$$\begin{aligned}
\sum_{i < j}^N \binom{N}{4}^{-1} \sum_{k < l < m < n} \phi_{klmn,ij} &= \sum_{i < j}^N \binom{N}{4}^{-1} \binom{N-2}{2} \bar{s}_{2,ij} \\
&= \sum_{i < j}^N \left\{ \frac{4! (N-4)!}{N!} \right\} \left\{ \frac{(N-2)!}{2! (N-4)!} \right\} \bar{s}_{2,ij} \\
&= \sum_{i < j}^N \left\{ \frac{12}{N(N-1)} \right\} \bar{s}_{2,ij}.
\end{aligned}$$

To derive the form of  $\mathbb{C}(U_N^*, U_N)$  given in the proof note that

$$\begin{aligned}
6 \binom{N}{4}^{-1} \binom{N-2}{2} \Delta_{2,N} &= 6 \frac{4! (N-4)!}{N!} \frac{(N-2)!}{2! (N-4)!} \Delta_{2,N} \\
&= \frac{72}{N(N-1)} \Delta_{2,N}.
\end{aligned}$$

## Calculation details for proof of Theorem 4

**Probability limit of concentrated Hessian:** The expression for  $H_{N,\beta\beta} + H_{N,\beta\mathbf{A}} V_N^{-1} H_{N,\beta\mathbf{A}}$ , the approximate Hessian of the concentrated log-likelihood given in (48), may be calculated

as follows

$$\begin{aligned}
& H_{N,\beta\beta} + H_{N,\beta\mathbf{A}} V_N^{-1} H_{N,\beta\mathbf{A}} \\
&= - \sum_{i=1}^N \sum_{j<i} p_{ij} (1 - p_{ij}) W_{ij} W'_{ij} \\
&\quad + \left( - \sum_{j \neq 1} p_{1j} (1 - p_{1j}) W_{1j} \quad \cdots \quad - \sum_{j \neq N} p_{Nj} (1 - p_{Nj}) W_{Nj} \right) \\
&\quad \times \text{diag} \left\{ \frac{1}{\sum_{j \neq 1} p_{1j} (1 - p_{1j})}, \dots, \frac{1}{\sum_{j \neq N} p_{Nj} (1 - p_{Nj})} \right\}' \\
&\quad \times \begin{pmatrix} - \sum_{j \neq 1} p_{1j} (1 - p_{1j}) W'_{1j} \\ \vdots \\ - \sum_{j \neq N} p_{Nj} (1 - p_{Nj}) W'_{Nj} \end{pmatrix} \\
&= - \sum_{i=1}^N \sum_{j<i} p_{ij} (1 - p_{ij}) W_{ij} W'_{ij} \\
&\quad \left( \frac{- \sum_{j \neq 1} p_{1j} (1 - p_{1j}) W_{1j}}{\sum_{j \neq 1} p_{1j} (1 - p_{1j})} \quad \cdots \quad \frac{- \sum_{j \neq N} p_{Nj} (1 - p_{Nj}) W_{Nj}}{\sum_{j \neq N} p_{Nj} (1 - p_{Nj})} \right) \begin{pmatrix} - \sum_{j \neq 1} p_{1j} (1 - p_{1j}) W'_{1j} \\ \vdots \\ - \sum_{j \neq N} p_{Nj} (1 - p_{Nj}) W'_{Nj} \end{pmatrix} \\
&= - \left\{ \sum_{i=1}^N \sum_{j<i} p_{ij} (1 - p_{ij}) W_{ij} W'_{ij} - \sum_{i=1}^N \frac{\left( \sum_{j \neq i} p_{ij} (1 - p_{ij}) W_{ij} \right) \left( \sum_{j \neq i} p_{ij} (1 - p_{ij}) W'_{ij} \right)'}{\sum_{j \neq i} p_{ij} (1 - p_{ij})} \right\}.
\end{aligned}$$

**Analysis of remainder term in (50):** Let  $f(v) = \frac{\exp(v)}{1+\exp(v)}$  be the logit function. To bound the third term in (50) I begin by calculating the derivative of  $f(v)(1-f(v))(1-2f(v)) = f(v) - 3f(v)^2 + 2f(v)^3$  with respect to  $v$ . Using the fact that  $f'(v) = f(v)(1-f(v))$  I get

$$\begin{aligned}
\frac{\partial}{\partial v} \{f(v)(1-f(v))(1-2f(v))\} &= f(v)(1-f(v)) - 6f(v)^2(1-f(v)) + 6f(v)^3(1-f(v)) \\
&= f(v)(1-f(v))(1-6f(v)+6f(v)^2) \\
&= f(v)(1-f(v))(1-6f(v)(1-f(v))).
\end{aligned}$$

Using condition (16) then gives

$$\sup_{1 \leq i, j \leq N} |p_{ij}(1-p_{ij})(1-6p_{ij}(1-p_{ij}))W_{ij}| \leq \frac{1}{4}(1-6\kappa(1-\kappa)) \times \sup_{w \in \mathbb{W}} |w|.$$

Expanding the fourth term in (50) I get

$$\begin{aligned}
&= \sum_{k=1}^N \sum_{l=1}^N \left( \hat{A}_k(\beta_0) - A_k(\beta_0) \right) \left( \hat{A}_l(\beta_0) - A_l(\beta_0) \right) \\
&\quad \times \left[ \sum_{i=1}^N \sum_{j < i} \frac{\partial^3}{\partial A_k \partial A_l \partial \mathbf{A}'} s_{\beta ij}(\beta_0, \bar{\mathbf{A}}(\beta_0)) \right] \\
&= - \sum_{k=1}^N \sum_{l \neq k} \left( \begin{array}{c} 0 \\ \vdots \\ \left( \hat{A}_k - A_k \right) \left( \hat{A}_l - A_l \right) p_{kl} (1 - p_{kl}) (1 - 6p_{kl} (1 - p_{kl})) W'_{kl} \\ \vdots \\ \left( \hat{A}_k - A_k \right) \left( \hat{A}_l - A_l \right) p_{kl} (1 - p_{kl}) (1 - 6p_{kl} (1 - p_{kl})) W'_{kl} \\ \vdots \\ 0 \end{array} \right)' \\
&= -2 \left( \begin{array}{c} \left( \hat{A}_1 - A_1 \right) \sum_{j \neq 1} \left( \hat{A}_j - A_j \right) p_{1j} (1 - p_{1j}) (1 - 6p_{1j} (1 - p_{1j})) W'_{1j} \\ \vdots \\ \left( \hat{A}_N - A_N \right) \sum_{j \neq N} \left( \hat{A}_j - A_j \right) p_{Nj} (1 - p_{Nj}) (1 - 6p_{Nj} (1 - p_{Nj})) W'_{1j} \end{array} \right)'.
\end{aligned}$$

Multiplying this by the  $N \times 1$  vector  $\hat{\mathbf{A}} - \mathbf{A}$  yields the  $K \times 1$  vector

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$$2 \sum_{i=1}^N \sum_{j \neq i} \left( \hat{A}_i - A_i \right)^2 \left( \hat{A}_j - A_j \right) (1 - p_{ij}) (1 - 6p_{ij} (1 - p_{ij})) W_{ij}$$

which gives (51) of the main text.

**Derivation of asymptotic bias:** To derive (52) it is convenient to proceed regressor by regressor. Observe that the  $k^{th}$  element of the third term appearing in (50) is, for  $k = 1, \dots, K$ ,

$$\frac{1}{2} \frac{1}{\sqrt{n}} \left[ \sum_{l=1}^N \left( \hat{A}_l(\beta_0) - A_l(\beta_0) \right) \sum_{i=1}^N \sum_{j < i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta ij}^{(k)}(\beta_0, \mathbf{A}(\beta_0)) \right] \left( \hat{\mathbf{A}}(\beta_0) - \mathbf{A}(\beta_0) \right) \quad (56)$$

The probability limit of (56) equals (52). To simplify (56) and, derive this limit, start by

observing that, for  $l = 1, \dots, N$ ,

$$\begin{aligned}
\sum_{i=1}^N \sum_{j < i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta_{ij}}(\beta_0, \mathbf{A}(\beta_0)) &= - \sum_{i=1}^N \sum_{j < i} p_{ij} (1 - p_{ij}) (1 - 2p_{ij}) W_{ij} T'_{ij} T_{l,ij} \\
&= - \frac{1}{2} \sum_{i=1}^N \sum_{j \neq i} p_{ij} (1 - p_{ij}) (1 - 2p_{ij}) W_{ij} T'_{ij} T_{l,ij} \\
&\quad - \left( p_{1l} (1 - p_{1l}) (1 - 2p_{1l}) W_{1l} \right. \\
&\quad \cdots \quad p_{l-1l} (1 - p_{l-1l}) (1 - 2p_{l-1l}) W_{l-1l} \\
&\quad \sum_{j \neq l} p_{lj} (1 - p_{lj}) (1 - 2p_{lj}) W_{lj} \\
&\quad p_{l+1l} (1 - p_{l+1l}) (1 - 2p_{l+1l}) W_{l+1l} \\
&\quad \left. \cdots \quad p_{Nl} (1 - p_{Nl}) (1 - 2p_{Nl}) W_{Nl} \right).
\end{aligned}$$

Next, using (31) from the proof of Lemma 6 and recalling that  $e_l$  is a conformable selection vector with a 1 in its  $l^{th}$  element and zeros elsewhere, gives

$$\hat{A}_l(\beta_0) - A_l(\beta_0) = -e'_l H_{N, \mathbf{A}\mathbf{A}}^{-1} \begin{pmatrix} D_{1+} - p_{1+} \\ \vdots \\ D_{N+} - p_{N+} \end{pmatrix} + o_p(1).$$

which allows the  $k^{th}$  element of the third term in (50) to be replaced with its asymptotic equivalent

$$\begin{aligned}
\frac{1}{2} \frac{1}{\sqrt{n}} \left[ \sum_{l=1}^N \begin{pmatrix} D_{1+} - p_{1+} \\ \vdots \\ D_{N+} - p_{N+} \end{pmatrix}' H_{N, \mathbf{A}\mathbf{A}}^{-1} e_l \left\{ \sum_{i=1}^N \sum_{j < i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta_{ij}}^{(k)}(\beta_0, \mathbf{A}(\beta_0)) \right\} \right. \\
\left. H_{N, \mathbf{A}\mathbf{A}}^{-1} \begin{pmatrix} D_{1+} - p_{1+} \\ \vdots \\ D_{N+} - p_{N+} \end{pmatrix} \right]. \quad (57)
\end{aligned}$$

Applying the trace operator to (57) and cycling elements yields

$$\frac{1}{2} \frac{1}{\sqrt{n}} \sum_{l=1}^N \text{Tr} \left( \left\{ \sum_{i=1}^N \sum_{j<i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta ij}^{(k)}(\beta_0, \mathbf{A}(\beta_0)) \right\} H_{N, \mathbf{A} \mathbf{A}}^{-1} \begin{pmatrix} D_{1+} - p_{1+} \\ \vdots \\ D_{N+} - p_{N+} \end{pmatrix} \begin{pmatrix} D_{1+} - p_{1+} \\ \vdots \\ D_{N+} - p_{N+} \end{pmatrix}' H_{N, \mathbf{A} \mathbf{A}}^{-1} e_l \right),$$

which, after taking expectations conditional on  $\mathbf{X}$  and  $\mathbf{A}_0$ , gives

$$-\frac{1}{2} \frac{1}{\sqrt{n}} \sum_{l=1}^N \text{Tr} \left( \left\{ \sum_{i=1}^N \sum_{j<i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta ij}^{(k)}(\beta_0, \mathbf{A}(\beta_0)) \right\} H_{N, \mathbf{A} \mathbf{A}}^{-1} e_l \right) \quad (58)$$

The difference between (57) and its expectation (58) is  $o_p(1)$ . To see this observe that the diagonal elements of the  $N \times N$  matrix

$$\left[ \sum_{i=1}^N \sum_{j<i} s_{\mathbf{A} ij}(\beta_0, \mathbf{A}(\beta_0)) \right] \left[ \sum_{i=1}^N \sum_{j<i} s_{\mathbf{A} ij}(\beta_0, \mathbf{A}(\beta_0)) \right]' = \begin{pmatrix} D_{1+} - p_{1+} \\ \vdots \\ D_{N+} - p_{N+} \end{pmatrix} \begin{pmatrix} D_{1+} - p_{1+} \\ \vdots \\ D_{N+} - p_{N+} \end{pmatrix}'. \quad (59)$$

consist of the terms  $(D_{i+} - p_{i+})^2$  for  $i = 1, \dots, N$ . Fix  $i$ , order the balance of units arbitrarily, and define  $l_{j|i} = (D_{ij} - p_{ij})(D_{i+} - p_{i+}) - p_{ij}(1 - p_{ij})$ ; note that  $\{l_{j|i}\}_{j=1}^\infty$  is a martingale difference sequence (with  $\mathbb{E}[l_{j|i} | l_{1|i}, \dots, l_{j-1|i}] = 0$  and bounded moments). A law of large numbers for martingale difference sequences therefore gives (recalling that a  $+$  denotes summation over the omitted subscript)

$$\frac{1}{N-1} (D_{i+} - p_{i+})^2 \xrightarrow{p} \lim_{N \rightarrow \infty} \left\{ \frac{\sum_{j \neq i} p_{ij} (1 - p_{ij})}{N-1} \right\}.$$

A similar argument can be used to characterize the probability limits of the off-diagonal elements of (59)

$$\frac{1}{N-1} (D_{i+} - p_{i+})(D_{k+} - p_{k+}) \xrightarrow{p} \lim_{N \rightarrow \infty} \left\{ \frac{p_{ik}(1 - p_{ik})}{N-1} \right\}.$$

Together these results imply that  $H_{N, \mathbf{A} \mathbf{A}}^{-1} \left[ \sum_{i=1}^N \sum_{j<i} s_{\mathbf{A} ij}(\beta_0, \mathbf{A}(\beta_0)) \right] \left[ \sum_{i=1}^N \sum_{j<i} s_{\mathbf{A} ij}(\beta_0, \mathbf{A}(\beta_0)) \right]' = -I_N + o_p(1)$  and hence (58).



To evaluate (58) it is convenient to be able to replace  $H_{N,\mathbf{A}\mathbf{A}}^{-1}$  with  $-V_N$ :

$$\begin{aligned}
& -\frac{1}{2\sqrt{n}} \sum_{l=1}^N \text{Tr} \left( \left\{ \sum_{i=1}^N \sum_{j<i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta ij}^{(k)}(\beta_0, \mathbf{A}(\beta_0)) \right\} H_{N,\mathbf{A}\mathbf{A}}^{-1} e_l \right) \\
&= \frac{1}{2\sqrt{n}} \sum_{l=1}^N \text{Tr} \left( \left[ \sum_{i=1}^N \sum_{j<i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta ij}^{(k)}(\beta_0, \mathbf{A}(\beta_0)) \right] V_N^{-1} e_l \right) \\
&+ \frac{1}{2\sqrt{n}} \sum_{l=1}^N \text{Tr} \left( \left[ \sum_{i=1}^N \sum_{j<i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta ij}^{(k)}(\beta_0, \mathbf{A}(\beta_0)) \right] (-H_{N,\mathbf{A}\mathbf{A}}^{-1} - Q_N) e_l \right) \\
&+ \frac{1}{2\sqrt{n}} \sum_{l=1}^N \text{Tr} \left( \left[ \sum_{i=1}^N \sum_{j<i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta ij}^{(k)}(\beta_0, \mathbf{A}(\beta_0)) \right] (Q_{N,\mathbf{A}\mathbf{A}} - V_N^{-1}) e_l \right). \quad (60)
\end{aligned}$$

The first term in (60) coincides with the  $k^{th}$  element of the bias expression given in the statement of the theorem. Evaluating this term yields

$$\begin{aligned}
& \frac{1}{2} \frac{1}{\sqrt{n}} \sum_{l=1}^N \text{Tr} \left( \left[ \sum_{i=1}^N \sum_{j<i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta ij}^{(k)}(\beta_0, \mathbf{A}(\beta_0)) \right] V_N^{-1} e_l \right) = \\
& \frac{1}{2} \frac{1}{\sqrt{n}} \sum_{l=1}^N \left[ \sum_{i=1}^N \sum_{j<i} \frac{\partial^2}{\partial A_l \partial \mathbf{A}'} s_{\beta ij}^{(k)}(\beta_0, \mathbf{A}(\beta_0)) \right] \\
& \quad \times \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \frac{1}{\sum_{j \neq l} p_{lj} (1 - p_{lj})} \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \\
& -\frac{1}{2} \frac{1}{\sqrt{n}} \sum_{l=1}^N \frac{\sum_{j \neq l} p_{lj} (1 - p_{lj}) (1 - 2p_{lj}) W_{k,lj}}{\sum_{j \neq l} p_{lj} (1 - p_{lj})}.
\end{aligned}$$

The second and third terms are asymptotically negligible. Equation (52) follows directly.