

• Supplementary File •

Distributed optimal consensus of multiagent systems with Markovian switching topologies: synchronous and asynchronous communications

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Appendix A Proof of Theorem 1

(1) Stability proof. Construct the following Lyapunov function $V = V_1 + V_2 + V_3$, where $V_1 = \mathbb{E}[\sum_{i \in I[1, N]} \xi_i^T P \xi_i]$, $V_2 = \mathbb{E}[\sum_{i \in I[1, N]} \sum_{j \in I[1, N]} \frac{(\mu_{ij} - \mu)^2}{4v_{ij}} + \frac{(\iota_{ij} - \iota)^2}{2\sigma_{ij}}]$ with $\mu > \{\frac{4}{\lambda_2}, \frac{2}{\delta}\}$ and ι being positive constant to be specified, and $V_3 = \mathbb{E}[\mu \sum_{i \in I[1, N]} \pi_i + \iota_1 \sum_{i \in I[1, N]} \pi_i^1]$ with ι_1 being positive constant to be specified.

Let $V_1^m = \mathbb{E}[\sum_{i \in I[1, N]} \xi_i^T P \xi_i \mathbf{1}_{\{\sigma(t)=m\}}]$, in which $\mathbf{1}_{\{\sigma(t)=m\}}$ represents the Dirac measure satisfying $\mathbf{1}_{\{\sigma(t)=m\}} = 1$ if $\sigma(t) = m$, and $\mathbf{1}_{\{\sigma(t)=m\}} = 0$ if $\sigma(t) \neq m$. In combination with (9) and Lemma 2, one has

$$\begin{aligned} dV_1^m &= \mathbb{E} \left[\sum_{i \in I[1, N]} \xi_i^T P \xi_i d\mathbf{1}_{\{\sigma(t)=m\}} \right] + \mathbb{E} \left[\sum_{i \in I[1, N]} 2\xi_i^T P \dot{\xi}_i \mathbf{1}_{\{\sigma(t)=m\}} \right] dt \\ &= \sum_{q \in I[1, s]} \lambda_{qm} V_1^q dt + \mathbb{E} \left[\left(\sum_{i \in I[1, N]} \xi_i^T (PA + A^T P) \xi_i + 2 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \mu_{ij} \xi_i^T P B F(\tilde{x}_i - \tilde{x}_j) \right. \right. \\ &\quad \left. \left. + 2 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \iota_{ij} \xi_i^T P B \text{sgn}[F(\tilde{x}_i - \tilde{x}_j)] \right. \right. \\ &\quad \left. \left. + 2 \sum_{i \in I[1, N]} \xi_i^T P B \left[\varphi_i(\tilde{x}_i) - \frac{1}{N} \sum_{j \in I[1, N]} \varphi_j(\tilde{x}_j) \right] \right) \mathbf{1}_{\{\sigma(t)=m\}} \right] dt. \end{aligned} \quad (\text{A1})$$

Based on $\lambda_{mm} + \sum_{q \in I[1, s], q \neq m} \lambda_{mq} = 0$, one can easily obtain

$$\sum_{m \in I[1, s]} \sum_{q \in I[1, s]} \lambda_{qm} V_1^q = 0. \quad (\text{A2})$$

By Young's inequality and $\zeta^T \text{sgn}(\zeta) = \|\zeta\|_1$, one obtains

$$\begin{aligned} &2 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \mu_{ij} \xi_i^T P B F(\tilde{x}_i - \tilde{x}_j) \\ &\leq -\frac{1}{2} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \mu_{ij} (\tilde{x}_i - \tilde{x}_j)^T P B B^T P (\tilde{x}_i - \tilde{x}_j) + 2 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \mu_{ij} e_i^T P B B^T P e_i, \end{aligned} \quad (\text{A3a})$$

and

$$\begin{aligned} &2 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \iota_{ij} \xi_i^T P B \text{sgn}[F(\tilde{x}_i - \tilde{x}_j)] \\ &= - \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \iota_{ij} (\xi_i - \xi_j)^T F^T \text{sgn}[F(\tilde{x}_i - \tilde{x}_j)] \\ &\leq - \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \iota_{ij} \|F(\tilde{x}_i - \tilde{x}_j)\|_1 + 2N \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \iota_{ij} \|F e_i\|_1. \end{aligned} \quad (\text{A3b})$$

Set $\varphi(\tilde{x}) = (\varphi_1^T(\tilde{x}_1), \dots, \varphi_N^T(\tilde{x}_N))^T$. Based on Assumption 4, one has

$$2 \sum_{i \in I[1, N]} \xi_i^T P B \left[\varphi_i(\tilde{x}_i) - \frac{1}{N} \sum_{j \in I[1, N]} \varphi_j(\tilde{x}_j) \right]$$

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$$\begin{aligned}
 &= 2\xi^T(M \otimes PB)\varphi(\tilde{x}) \leq 2\|(M \otimes B^T P)\xi\|_1 \|(M \otimes I_n)\varphi(\tilde{x})\|_\infty \\
 &\leq \varpi(N-1) \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F(\tilde{x}_i - \tilde{x}_j)\|_1 + 2\varpi(N-1) \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F e_i\|_1,
 \end{aligned} \tag{A4}$$

where ϖ represents the supremum of $\|(M \otimes I_n)\varphi(\tilde{x})\|_\infty$.

Substituting (A2), (A3), and (A4) into (A1) and combining with $V_1 = \sum_{m \in I[1, s]} V_1^m$ yield

$$\begin{aligned}
 \dot{V}_1 \leq & \mathbb{E} \left[\sum_{i \in I[1, N]} \xi_i^T (PA + A^T P) \xi_i - \frac{1}{2} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} \mu_{ij} a_{ij}^{\sigma(t)} (\tilde{x}_i - \tilde{x}_j)^T G (\tilde{x}_i - \tilde{x}_j) \right. \\
 & + 2 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} \mu_{ij} a_{ij}^{\sigma(t)} e_i^T G e_i - \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \iota_{ij} \|F(\tilde{x}_i - \tilde{x}_j)\|_1 \\
 & + 2N \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \iota_{ij} \|F e_i\|_1 + \varpi(N-1) \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F(\tilde{x}_i - \tilde{x}_j)\|_1 \\
 & \left. + 2\varpi(N-1) \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F e_i\|_1 \right].
 \end{aligned} \tag{A5}$$

Let $V_2^m = \mathbb{E}[(\sum_{i \in I[1, N]} \sum_{j \in I[1, N]} \frac{(\mu_{ij} - \mu)^2}{4v_{ij}} + \frac{(\iota_{ij} - \iota)^2}{2o_{ij}}) \mathbf{1}_{\{\sigma(t)=m\}}]$. In combination with (3b) and (3c), one has

$$\begin{aligned}
 dV_2^m = & \mathbb{E} \left[\sum_{i \in I[1, N]} \sum_{j \in I[1, N]} \left(\frac{(\mu_{ij} - \mu)^2}{4v_{ij}} + \frac{(\iota_{ij} - \iota)^2}{2o_{ij}} \right) d\mathbf{1}_{\sigma(t)=m} \right] \\
 & + \mathbb{E} \left[\left(\sum_{i \in I[1, N]} \sum_{j \in I[1, N]} \frac{\mu_{ij} - \mu}{2} a_{ij}^{\sigma(t)} \|F(\tilde{x}_i - \tilde{x}_j)\|_2^2 \right. \right. \\
 & \left. \left. + \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} (\iota_{ij} - \iota) a_{ij}^{\sigma(t)} \|F(\tilde{x}_i - \tilde{x}_j)\|_1 \right) \mathbf{1}_{\{\sigma(t)=m\}} \right] dt.
 \end{aligned} \tag{A6}$$

By $x_i - x_j = \xi_i - \xi_j + e_i - e_j$ and Young's inequality, one obtains

$$\begin{aligned}
 & -\frac{\mu}{4} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} (\tilde{x}_i - \tilde{x}_j)^T G (\tilde{x}_i - \tilde{x}_j) \\
 & \leq -\frac{\mu}{8} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} (\xi_i - \xi_j)^T G (\xi_i - \xi_j) + \mu \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} e_i^T G e_i.
 \end{aligned} \tag{A7}$$

Substituting (A7) into (A6) and combining with $V_2 = \sum_{m \in I[1, s]} V_2^m$ yield

$$\begin{aligned}
 \dot{V}_2 \leq & \mathbb{E} \left[\sum_{i \in I[1, N]} \sum_{j \in I[1, N]} \frac{2\mu_{ij} - \mu}{4} a_{ij}^{\sigma(t)} \|F(\tilde{x}_i - \tilde{x}_j)\|_2^2 + \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} (\iota_{ij} - \iota) a_{ij}^{\sigma(t)} \|F(\tilde{x}_i - \tilde{x}_j)\|_1 \right. \\
 & \left. - \frac{\mu}{8} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} (\xi_i - \xi_j)^T G (\xi_i - \xi_j) + \mu \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} e_i^T G e_i \right].
 \end{aligned} \tag{A8}$$

Similar to (A5) and (A8), and combining with (7b) and (7d), one has

$$\begin{aligned}
 \dot{V}_3 = & \mathbb{E} \left[\frac{\mu\gamma_2}{4} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} (\tilde{x}_i - \tilde{x}_j)^T G (\tilde{x}_i - \tilde{x}_j) \right. \\
 & - \mu\gamma_2 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} (1 + \delta\mu_{ij}) a_{ij}^{\sigma(t)} e_i^T G e_i - \mu\gamma_3 \sum_{i \in I[1, N]} \pi_i \\
 & \left. + \iota_1 \sum_{i \in I[1, N]} \left(\gamma_2^1 \left\{ \delta_2 \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F(\tilde{x}_i - \tilde{x}_j)\|_1 - \sum_{j \in I[1, N]} (1 + \delta^1 \iota_{ij}) a_{ij}^{\sigma(t)} \|F e_i\|_1 \right\} - \gamma_3^1 \pi_i^1 \right) \right].
 \end{aligned} \tag{A9}$$

Then, combining \dot{V}_1 in (A5), \dot{V}_2 in (A8), \dot{V}_3 in (A9) and triggering mechanism in (6), we obtain

$$\begin{aligned}
 \dot{V} \leq & \mathbb{E} \left[\sum_{i \in I[1, N]} \xi_i^T (PA + A^T P) \xi_i - \frac{\mu}{4} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \xi_i^T G (\xi_i - \xi_j) \right. \\
 & + \mu[(1 - \gamma_2)\gamma_1 - \gamma_3] \sum_{i \in I[1, N]} \pi_i + [2\varpi N \delta_2 + \varpi(N-1) - \iota] \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F(\tilde{x}_i - \tilde{x}_j)\|_1 \\
 & \left. + [(2\varpi N - \iota_1 \gamma_2^1) \gamma_1^1 - \iota_1 \gamma_3] \sum_{i \in I[1, N]} \pi_i^1 \right],
 \end{aligned} \tag{A10}$$

where $\iota > 2\varpi N \delta_2 + \varpi(N-1)$ and $(2\varpi N - \iota_1 \gamma_2^1) \gamma_1^1 - \iota_1 \gamma_3 < 0$ can be ensured by selecting ι_1 .

When $t \in [t_k^i, t_{k+1}^i)$, it follows from triggering mechanism in (15) that $\dot{\pi}_i \geq -(\gamma_1 \gamma_2 + \gamma_3) \pi_i$, which implies that $\pi_i \geq \pi_i(t_k^i) e^{-(\gamma_1 \gamma_2 + \gamma_3)(t - t_k^i)} \geq \dots \geq \pi_i(0) e^{-(\gamma_1 \gamma_2 + \gamma_3)t} > 0$. Similarly, $\pi_i^1 > \pi_i^1(0) e^{-(\gamma_1^1 \gamma_2^1 + \gamma_3^1)t} > 0$ can be ensured.

Since $\gamma_3 > (1 - \gamma_2)\gamma_1$, Eq. (A10) can be written as

$$\dot{V} \leq \xi^T \left[I_N \otimes (AP + PA^T) - \frac{\mu}{4} \mathbb{E}[\mathcal{L}^{\sigma(t)}] \otimes G \right] \xi. \quad (\text{A11})$$

Based on Assumptions 1 and 3 and Lemma 1, there must exist an orthogonal matrix \mathcal{O} such that $\mathcal{O}^T \mathbb{E}[\mathcal{L}^{\sigma(t)}] \mathcal{O} = \text{diag}\{0, \lambda_2, \dots, \lambda_N\}$ ascending. Set $\bar{\xi} = [\bar{\xi}_1^T, \dots, \bar{\xi}_N^T]^T = (\mathcal{O}^T \otimes I_n) \xi$, from which we can obtain $\bar{\xi}_1 = \mathbf{0}$. Then, Eq. (A11) can be written as

$$\dot{V} \leq \sum_{i \in I[2, N]} \bar{\xi}_i^T \left(AP + PA^T - \frac{\mu \lambda_i}{4} P B B^T P \right) \bar{\xi}_i \leq - \sum_{i \in I[2, N]} \bar{\xi}_i^T \bar{\xi}_i \quad (\text{A12})$$

which implies that ξ_i is stable. Therefore, $x_i \rightarrow \frac{1}{N} \sum_{i \in I[1, N]} x_i \triangleq \bar{x}$ can be obtained as $t \rightarrow \infty$.

(2) Optimization proof. Based on Assumption 5 and extreme value theorem, we can obtain that $\sum_{i \in I[1, N]} f_i(\bar{x})$ has a lower bound. Therefore, one obtains that $\frac{d \sum_{i \in I[1, N]} f_i(\bar{x})}{dt} = \mathbf{0}$. Then, consider the following formula:

$$\begin{aligned} & \lim_{t \rightarrow \infty} \left\{ \frac{d \sum_{i \in I[1, N]} f_i(\bar{x})}{dt} - \frac{1}{N} \left(\frac{\partial \sum_{i \in I[1, N]} f_i(\bar{x})}{\partial \bar{x}} \right)^T \mathcal{R}^{-1} \left(\frac{\partial \sum_{i \in I[1, N]} f_i(\bar{x})}{\partial \bar{x}} \right) \right\} \\ &= \lim_{t \rightarrow \infty} \left\{ \left(\frac{\partial \sum_{i \in I[1, N]} f_i(\bar{x})}{\partial \bar{x}} \right)^T \left[\frac{d\bar{x}}{dt} - \frac{1}{N} \mathcal{R}^{-1} \left(\frac{\partial \sum_{i \in I[1, N]} f_i(\bar{x})}{\partial \bar{x}} \right) \right] \right\}. \end{aligned} \quad (\text{A13})$$

Since the following formula is satisfied:

$$\begin{aligned} & \lim_{t \rightarrow \infty} \left\{ \frac{d\bar{x}}{dt} - \frac{1}{N} \mathcal{R}^{-1} \left(\frac{\partial \sum_{i \in I[1, N]} f_i(\bar{x})}{\partial \bar{x}} \right) \right\} \\ &= \lim_{t \rightarrow \infty} \left\{ \frac{1}{N} \sum_{i \in I[1, N]} [Ax_i + B\varphi(x_i)] - \frac{1}{N} \mathcal{R}^{-1} \sum_{i \in I[1, N]} \mathcal{R}A[\bar{x} + B\varphi(\bar{x})] \right\} = \mathbf{0}. \end{aligned} \quad (\text{A14})$$

Combining (A13) and (A14), and $\lim_{t \rightarrow \infty} \frac{d \sum_{i \in I[1, N]} f_i(\bar{x})}{dt} = \mathbf{0}$ one obtains

$$\lim_{t \rightarrow \infty} \left(\frac{\partial \sum_{i \in I[1, N]} f_i(\bar{x})}{\partial \bar{x}} \right)^T \mathcal{R}^{-1} \left(\frac{\partial \sum_{i \in I[1, N]} f_i(\bar{x})}{\partial \bar{x}} \right) = \mathbf{0}.$$

Since matrix \mathcal{R} is nonsingular, one can further obtain $\lim_{t \rightarrow \infty} \frac{\partial \sum_{i \in I[1, N]} f_i(\bar{x})}{\partial \bar{x}} = \mathbf{0}$, which implies \bar{x} is the optimal state by Lemma 3.

(3) Feasibility proof. Here, the proof by contradiction method is adopted. Suppose agent i exhibits Zeno behavior, that is, $\lim_{t \rightarrow \infty} T_k^i = 0$ with $T_k^i \triangleq t_{k+1}^i - t_k^i$, $\lim_{k \rightarrow \infty} t_k^i = t_\infty^i < \infty$. At triggering instant t_{k+1}^i , at least one of the following inequations holds:

$$\sum_{j \in I[1, N]} (1 + \delta \mu_{ij}) a_{ij}^{\sigma(t)} \|F e_i(t_{k+1}^i)\|_2^2 \geq \gamma_1 \pi_1(t_{k+1}^i) \geq \gamma_1 \pi_1(0) e^{-(\gamma_1 \gamma_2 + \gamma_3) t_{k+1}^i}$$

and

$$\sum_{j \in I[1, N]} (1 + \delta' \nu_{ij}) a_{ij}^{\sigma(t)} \|F e_i\|_1 \geq \gamma_1^1 \pi_1^1(t_{k+1}^i) \geq \gamma_1^1 \pi_1^1(0) e^{-(\gamma_1^1 \gamma_2^1 + \gamma_3^1) t_{k+1}^i}.$$

Therefore, $0 = \lim_{t \rightarrow \infty} \sum_{j \in I[1, N]} (1 + \delta \mu_{ij}) a_{ij}^{\sigma(t)} \|F e_i(t_{k+1}^i)\|_2^2 \geq \gamma_1 \pi_1(0) e^{-(\gamma_1 \gamma_2 + \gamma_3) t_{k+1}^i} > 0$ or $0 = \lim_{t \rightarrow \infty} \sum_{j \in I[1, N]} (1 + \delta' \nu_{ij}) a_{ij}^{\sigma(t)} \|F e_i(t_{k+1}^i)\|_1 \geq \gamma_1^1 \pi_1^1(0) e^{-(\gamma_1^1 \gamma_2^1 + \gamma_3^1) t_{k+1}^i} > 0$, which is obviously contradictory. Thus, no Zeno behavior is exhibited under the proposed triggering mechanism in (6). The proof is completed.

Appendix B Proof of Theorem 2

Stability proof. Construct the following Lyapunov function $V = V_1 + V_2 + V_3$, where $V_1 = \mathbb{E}[\sum_{i \in I[1, N]} \xi_i^T P \xi_i]$, $V_2 = \mathbb{E}[\sum_{i \in I[1, N]} \sum_{j \in I[1, N]} \left(\frac{(d_{ij} - d)^2}{4m_{ij}} + \frac{(c_{ij} - c)^2}{2\tau_{ij}} \right)]$ with $d > \left\{ \frac{4}{\lambda_2}, \frac{2}{\varpi} \right\}$ and c being positive constant to be specified, and $V_3 = \mathbb{E}[d \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \pi_{ij} + c_1 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \pi_{ij}^1]$ with c_1 being positive constant to be specified. Similar to (A5), calculate the derivative of V_1 along (16) as

$$\begin{aligned} \dot{V}_1 \leq & \mathbb{E} \left[\sum_{i \in I[1, N]} \xi_i^T (PA + A^T P) \xi_i - \frac{1}{2} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} d_{ij} a_{ij}^{\sigma(t)} (\tilde{x}_{ij} - \tilde{x}_{ji})^T G (\tilde{x}_{ij} - \tilde{x}_{ji}) \right. \\ & + 2 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} d_{ij} a_{ij}^{\sigma(t)} e_{ij}^T G e_{ij} - \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} c_{ij} \|F(\tilde{x}_{ij} - \tilde{x}_{ji})\|_1 \\ & + 2N \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} c_{ij} \|F e_{ij}\|_1 + \varpi(N-1) \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F(\tilde{x}_{ij} - \tilde{x}_{ji})\|_1 \\ & \left. + 2\varpi(N-1) \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F e_{ij}\|_1 \right]. \end{aligned} \quad (\text{B1})$$

Similar to (A8), one can obtain

$$\begin{aligned}
 \dot{V}_2 \leq & \mathbb{E} \left[\sum_{i \in I[1, N]} \sum_{j \in I[1, N]} \frac{d_{ij}}{2} a_{ij}^{\sigma(t)} (\bar{x}_{ij} - \bar{x}_{ji})^T G (\bar{x}_{ij} - \bar{x}_{ji}) - \frac{d}{4} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} (\bar{x}_{ij} - \bar{x}_{ji})^T G (\bar{x}_{ij} - \bar{x}_{ji}) \right. \\
 & - \frac{d}{8} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} (\xi_i - \xi_j)^T G (\xi_i - \xi_j) + d \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} e_{ij}^T G e_{ij} \\
 & \left. + \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} (c_{ij} - c) a_{ij}^{\sigma(t)} \|F(\bar{x}_{ij} - \bar{x}_{ji})\|_1 \right]. \tag{B2}
 \end{aligned}$$

Similar to (A9), calculate the derivative of V_3 along (14b) and (14d)

$$\begin{aligned}
 \dot{V}_3 = & \mathbb{E} \left[\frac{d\bar{\gamma}_2}{4} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F(\bar{x}_{ij} - \bar{x}_{ji})\|_2^2 - d\bar{\gamma}_2 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} (1 + \sigma d_{ij}) a_{ij}^{\sigma(t)} \|F e_{ij}\|_2^2 \right. \\
 & - d\bar{\gamma}_3 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \pi_{ij} + c_1 \bar{\gamma}_2^1 \sigma_2 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F(\bar{x}_{ij} - \bar{x}_{ji})\|_1 \\
 & \left. - c_1 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} (1 + \sigma^1 c_{ij}) \|F e_{ij}\|_1 - c_1 \bar{\gamma}_3^1 \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \pi_{ij}^1 \right]. \tag{B3}
 \end{aligned}$$

Combining (B1)–(B3) and triggering mechanism in (13) obtain

$$\begin{aligned}
 \dot{V} \leq & \mathbb{E} \left[\sum_{i \in I[1, N]} \xi_i^T (PA + A^T P) \xi_i - \frac{d}{4} \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \xi_i^T G (\xi_i - \xi_j) + d[(1 - \bar{\gamma}_2)\bar{\gamma}_1 - \bar{\gamma}_3] \right. \\
 & \otimes \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \pi_{ij} + [2\varpi N \sigma_2 + \varpi(N - 1) - c] \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \|F(\bar{x}_{ij} - \bar{x}_{ji})\|_1 \\
 & \left. + [(2\varpi N - c_1 \bar{\gamma}_2^1)\bar{\gamma}_1^1 - c_1 \bar{\gamma}_3] \sum_{i \in I[1, N]} \sum_{j \in I[1, N]} a_{ij}^{\sigma(t)} \pi_{ij}^1 \right], \tag{B4}
 \end{aligned}$$

where $c > 2\varpi N \sigma_2 + \varpi(N - 1)$ and $(2\varpi N - c_1 \bar{\gamma}_2^1)\bar{\gamma}_1^1 - c_1 \bar{\gamma}_3^1 < 0$ can be ensured by selecting c_1 .

Similar to (A11), Eq. (B4) can be written as

$$\dot{V} \leq \mathbb{E} \left[\xi^T \left[I_N \otimes (AP + PA^T) - \frac{\mu}{4} \mathcal{L}^{\sigma(t)} \otimes G \right] \xi \right]. \tag{B5}$$

Then, similar to Theorem 1, one can obtain $\lim_{t \rightarrow \infty} \xi = \mathbf{0}$.

The proof of optimization and feasibility is the same as Theorem 1, which is omitted here.