Information Consensus for Distributed Multi-Target Tracking Supplementary Materials

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1. JPDAF: Information Form

The JPDAF estimation and covariance update equation are given in the following (Eqns. (15-16) in the main paper) which are then converted to the equivalent information form (Eqns. (18-19) in the main paper) below. The time index t has been dropped for simplicity.

$$\begin{split} \mathbf{K}^{j} &= \mathbf{P}^{j-}\mathbf{H}^{jT}\mathbf{S}^{j-1} \\ &= \mathbf{P}^{j-}\mathbf{H}^{jT}\left(\mathbf{H}^{j}\mathbf{P}^{j-}\mathbf{H}^{jT} + \mathbf{R}^{j}\right)^{-1} \\ &= \mathbf{P}^{j-}\mathbf{H}^{jT}\left(\mathbf{R}^{j-1} - \mathbf{R}^{j-1}\mathbf{H}^{j}\left((\mathbf{P}^{j-})^{-1} + \mathbf{H}^{jT}\mathbf{R}^{j-1}\mathbf{H}^{j}\right)^{-1}\mathbf{H}^{jT}\mathbf{R}^{j-1}\right) \quad \text{(using Matrix Inversion Lemma)} \\ &= \left(\mathbf{P}^{j-} - \mathbf{P}^{j-}\mathbf{H}^{jT}\mathbf{R}^{j-1}\mathbf{H}^{j}\left((\mathbf{P}^{j-})^{-1} + \mathbf{H}^{jT}\mathbf{R}^{j-1}\mathbf{H}^{j}\right)^{-1}\right)\mathbf{H}^{jT}\mathbf{R}^{j-1} \\ &= \left(\mathbf{P}^{j-}\left((\mathbf{P}^{j-})^{-1} + \mathbf{H}^{jT}\mathbf{R}^{j-1}\mathbf{H}^{j}\right) - \mathbf{P}^{j-}\mathbf{H}^{jT}\mathbf{R}^{j-1}\mathbf{H}^{j}\right)\left((\mathbf{P}^{j-})^{-1} + \mathbf{H}^{jT}\mathbf{R}^{j-1}\mathbf{H}^{j}\right)^{-1}\mathbf{H}^{jT}\mathbf{R}^{j-1} \\ &= \left(\mathbf{I}_{p} + \mathbf{P}^{j-}\mathbf{H}^{jT}\mathbf{R}^{j-1}\mathbf{H}^{j} - \mathbf{P}^{j-}\mathbf{H}^{jT}\mathbf{R}^{j-1}\mathbf{H}^{j}\right)\left((\mathbf{P}^{j-})^{-1} + \mathbf{H}^{jT}\mathbf{R}^{j-1}\mathbf{H}^{j}\right)^{-1}\mathbf{H}^{jT}\mathbf{R}^{j-1} \\ &= \left((\mathbf{P}^{j-})^{-1} + \mathbf{H}^{jT}\mathbf{R}^{j-1}\mathbf{H}^{j}\right)^{-1}\mathbf{H}^{jT}\mathbf{R}^{j-1} \end{split}$$

$$\tilde{\mathbf{y}}^j = \mathbf{y}^j - (1 - \beta^{j0}) \mathbf{H}^j \hat{\mathbf{x}}^{j-}$$
(2)

$$\hat{\mathbf{x}}^{j+} = \hat{\mathbf{x}}^{j-} + \mathbf{K}^{j} \tilde{\mathbf{y}}^{j}
= \hat{\mathbf{x}}^{j-} + \left((\mathbf{P}^{j-})^{-1} + \mathbf{H}^{j}^{T} \mathbf{R}^{j-1} \mathbf{H}^{j} \right)^{-1} \mathbf{H}^{j}^{T} \mathbf{R}^{j-1} \left(\mathbf{y}^{j} - (1 - \beta^{j0}) \mathbf{H}^{j} \hat{\mathbf{x}}^{j-} \right)
= \hat{\mathbf{x}}^{j-} + \left(\mathbf{J}^{j-} + \mathbf{U}^{j} \right)^{-1} \left(\mathbf{u}^{j} - (1 - \beta^{j0}) \mathbf{U}^{j} \hat{\mathbf{x}}^{j-} \right)
= \left(\mathbf{J}^{j-} + \mathbf{U}^{j} \right)^{-1} \left(\mathbf{J}^{j-} \hat{\mathbf{x}}^{j-} + \mathbf{U}^{j} \hat{\mathbf{x}}^{j-} + \mathbf{u}^{j} - (1 - \beta^{j0}) \mathbf{U}^{j} \hat{\mathbf{x}}^{j-} \right)
= \left(\mathbf{J}^{j-} + \mathbf{U}^{j} \right)^{-1} \left(\mathbf{J}^{j-} \hat{\mathbf{x}}^{j-} + \mathbf{u}^{j} + \beta^{j0} \mathbf{U}^{j} \hat{\mathbf{x}}^{j-} \right)$$
(3)

$$\mathbf{P}^{j+} = \mathbf{P}^{j-} - (1 - \beta^{j0}) \mathbf{K}^{j} \mathbf{S}^{j} \mathbf{K}^{jT} + \mathbf{K}^{j} \tilde{\mathbf{P}}^{j} \mathbf{K}^{jT}$$

$$= \mathbf{P}^{j-} - \mathbf{K}^{j} \left((1 - \beta^{j0}) \mathbf{S}^{j} - \tilde{\mathbf{P}}^{j} \right) \mathbf{K}^{jT}$$
(4)

where,

$$\tilde{\mathbf{P}}^{j} = \left(\sum_{n=1}^{l} \beta^{jn} \tilde{\mathbf{z}}^{jn} \left(\tilde{\mathbf{z}}^{jn}\right)^{T}\right) - \tilde{\mathbf{y}}^{j} \left(\tilde{\mathbf{y}}^{j}\right)^{T}$$
(5)

Let,

$$\mathbf{C}^{j} = (1 - \beta^{j0})\mathbf{S}^{j} - \tilde{\mathbf{P}}^{j} \tag{6}$$

Thus, using matrix inversion lemma and by definition of $\mathbf{J}^{j+} = (\mathbf{P}^{j+})^{-1}$ and $\mathbf{J}^{j-} = (\mathbf{P}^{j-})^{-1}$ we get,

$$\mathbf{J}^{j+} = \mathbf{J}^{j-} + \mathbf{J}^{j-} \mathbf{K}^{j} \left(\left(\mathbf{C}^{j} \right)^{-1} - \mathbf{K}^{jT} \mathbf{J}^{j-} \mathbf{K}^{j} \right)^{-1} \mathbf{K}^{jT} \mathbf{J}^{j-}$$

$$= \mathbf{J}^{j-} + \mathbf{G}^{j}$$
(7)

where

$$\mathbf{G}^{j} = \mathbf{J}^{j-} \mathbf{K}^{j} \left(\left(\mathbf{C}^{j} \right)^{-1} - \mathbf{K}^{jT} \mathbf{J}^{j-} \mathbf{K}^{j} \right)^{-1} \mathbf{K}^{jT} \mathbf{J}^{j-}.$$
 (8)

Thus, Equ. (3) and (7) are the JPDAF estimation equations in the information form. These results i.e., Eqns. (3,7,8,6) are shown as results in the main paper as Eqns. (18-21).

2. Comparison of algorithms in innovation form: Results

For simplicity, here we will drop the target superscript. Define the single step consensus operation A on a value a_i as,

$$\mathcal{A}(a_i) = a_i + \epsilon \sum_{i' \in \mathcal{N}_i} (a_{i'} - a_i) \tag{9}$$

Also,

$$\mathbf{B}_{i} = \sum_{i' \in \mathcal{N}_{i} \cup \{i\}} \mathbf{U}_{i'} \tag{10}$$

$$\mathbf{b}_{i} = \sum_{i' \in \mathcal{N}_{i} \cup \{i\}} \mathbf{u}_{i'} \tag{11}$$

Note that in general, $\mathbf{B}_i \neq N_C \mathcal{A}(\mathbf{U}_i)$ and $\mathbf{b}_i \neq N_C \mathcal{A}(\mathbf{u}_i)$. We can get the following results for a single consensus step, which will be derived later.

2.1. KCF

(Eqns. (4-5) in the main paper)

$$\hat{\mathbf{x}}_{i}^{+} = \hat{\mathbf{x}}_{i}^{-} + (\mathbf{J}_{i}^{-} + \mathbf{B}_{i})^{-1} (\mathbf{b}_{i} - \mathbf{B}_{i} \hat{\mathbf{x}}_{i}^{-}) + \frac{\epsilon}{1 + ||(\mathbf{J}_{i}^{-})^{-1}||} (\mathbf{J}_{i}^{-})^{-1} \sum_{i' \in \mathcal{N}_{i}} (\hat{\mathbf{x}}_{i'}^{-} - \hat{\mathbf{x}}_{i}^{-})$$
(12)

$$\mathbf{J}_{i}^{+} = \mathbf{J}_{i}^{-} + \mathbf{B}_{i} \tag{13}$$

2.2. ICF

(Eqn. (40-41 in the main paper)

$$\hat{\mathbf{x}}_{i}^{+} = \hat{\mathbf{x}}_{i}^{-} + \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i})\right)^{-1} \left(\mathcal{A}(\mathbf{u}_{i}) - \mathcal{A}(\mathbf{U}_{i})\hat{\mathbf{x}}_{i}^{-} + \epsilon \sum_{i' \in \mathcal{N}_{i}} \frac{1}{N_{C}} \mathbf{J}_{i'}^{-} \left(\hat{\mathbf{x}}_{i'}^{-} - \hat{\mathbf{x}}_{i}^{-}\right)\right)$$
(14)

$$\mathbf{J}_{i}^{+} = N_{C} \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i}) \right)$$
 (15)

2.3. MTIC

(Eqn. (42-43) in the main paper)

$$\hat{\mathbf{x}}_{i}^{+} = \hat{\mathbf{x}}_{i}^{-} + \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i}) \right)^{-1} \left(\mathcal{A}(\mathbf{u}_{i}) - \mathcal{A}(\mathbf{U}_{i})\hat{\mathbf{x}}_{i}^{-} + \epsilon \sum_{i' \in \mathcal{N}_{i}} \frac{\mathbf{J}_{i'}^{-}}{N_{C}} \left(\hat{\mathbf{x}}_{i'}^{-} - \hat{\mathbf{x}}_{i}^{-} \right) + \mathcal{A}(\beta_{i0}\mathbf{U}_{i}\hat{\mathbf{x}}_{i}^{-}) \right)$$
(16)

$$\mathbf{J}_{i}^{+} = N_{C} \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i}) \right)$$
(17)

3. Proofs of the results:

3.1. ICF

For a single consensus step, ICF estimate equations (Eqn. (9) in the main paper) can be written with the $\mathcal{A}()$ notation as,

$$\hat{\mathbf{x}}_{i}^{+} = \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i}) \right)^{-1} \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}\hat{\mathbf{x}}_{i}^{-}) + \mathcal{A}(\mathbf{u}_{i}) \right)$$
(18)

$$\mathbf{J}_{i}^{+} = N_{C} \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i}) \right)$$
(19)

By adding and subtracting $\hat{\mathbf{x}}_i^-$ in RHS of (18) and denoting $\mathbf{F}_i = \frac{\mathbf{J}_i^-}{N_C}$, we get,

$$\hat{\mathbf{x}}_{i}^{+} = \hat{\mathbf{x}}_{i}^{-} + (\mathcal{A}(\mathbf{F}_{i}) + \mathcal{A}(\mathbf{U}_{i}))^{-1} \left(\mathcal{A}(\mathbf{F}_{i}\hat{\mathbf{x}}_{i}^{-}) + \mathcal{A}(\mathbf{u}_{i}) - (\mathcal{A}(\mathbf{F}_{i}) + \mathcal{A}(\mathbf{U}_{i})) \hat{\mathbf{x}}_{i}^{-} \right)$$
(20)

$$= \hat{\mathbf{x}}_{i}^{-} + (\mathcal{A}(\mathbf{F}_{i}) + \mathcal{A}(\mathbf{U}_{i}))^{-1} \left(\mathcal{A}(\mathbf{u}_{i}) - \mathcal{A}(\mathbf{U}_{i})\hat{\mathbf{x}}_{i}^{-} + \mathcal{A}(\mathbf{F}_{i}\hat{\mathbf{x}}_{i}^{-}) - \mathcal{A}(\mathbf{F}_{i})\hat{\mathbf{x}}_{i}^{-}\right)$$
(21)

Now,

$$\mathcal{A}(\mathbf{F}_{i}\hat{\mathbf{x}}_{i}^{-}) - \mathcal{A}(\mathbf{F}_{i})x_{i}^{-} = \mathbf{F}_{i}\hat{\mathbf{x}}_{i}^{-} + \epsilon \sum_{i' \in \mathcal{N}_{i}} \left(\mathbf{F}_{i}'\hat{\mathbf{x}}_{i'}^{-} - \mathbf{F}_{i}\hat{\mathbf{x}}_{i}^{-}\right) - \mathbf{F}_{i}\hat{\mathbf{x}}_{i}^{-} - \epsilon \sum_{i' \in \mathcal{N}_{i}} \left(\mathbf{F}_{i'}\hat{\mathbf{x}}_{i}^{-} - \mathbf{F}_{i}\hat{\mathbf{x}}_{i}^{-}\right)$$
(22)

$$= \epsilon \sum_{i' \in \mathcal{N}_i} \left(\mathbf{F}_{i'} \hat{\mathbf{x}}_{i'}^- - \mathbf{F}_i \hat{\mathbf{x}}_i^- \right) - \epsilon \sum_{i' \in \mathcal{N}_i} \left(\mathbf{F}_{i'} \hat{\mathbf{x}}_i^- - \mathbf{F}_i \hat{\mathbf{x}}_i^- \right)$$
(23)

$$= \epsilon \sum_{i' \in \mathcal{N}_i} \left(\mathbf{F}_{i'} \hat{\mathbf{x}}_{i'}^- - \mathbf{F}_i \hat{\mathbf{x}}_i^- - \mathbf{F}_{i'} \hat{\mathbf{x}}_i^- + \mathbf{F}_i \hat{\mathbf{x}}_i^- \right)$$
(24)

$$= \epsilon \sum_{i' \in \mathcal{N}_i} \mathbf{F}_{i'} \left(\hat{\mathbf{x}}_{i'}^- - \hat{\mathbf{x}}_i^- \right)$$
 (25)

Using this result in (21), we get,

$$\hat{\mathbf{x}}_{i}^{+} = \hat{\mathbf{x}}_{i}^{-} + (\mathcal{A}(\mathbf{F}_{i}) + \mathcal{A}(\mathbf{U}_{i}))^{-1} \left(\mathcal{A}(\mathbf{u}_{i}) - \mathcal{A}(\mathbf{U}_{i})\hat{\mathbf{x}}_{i}^{-} + \epsilon \sum_{i' \in \mathcal{N}_{i}} \mathbf{F}_{i'} \left(\hat{\mathbf{x}}_{i'}^{-} - \hat{\mathbf{x}}_{i}^{-} \right) \right)$$

$$= \hat{\mathbf{x}}_{i}^{-} + \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i}) \right)^{-1} \left(\mathcal{A}(\mathbf{u}_{i}) - \mathcal{A}(\mathbf{U}_{i})\hat{\mathbf{x}}_{i}^{-} + \epsilon \sum_{i' \in \mathcal{N}_{i}} \frac{\mathbf{J}_{i'}^{-}}{N_{C}} \left(\hat{\mathbf{x}}_{i'}^{-} - \hat{\mathbf{x}}_{i}^{-} \right) \right) \tag{26}$$

Thus, we have the ICF equations in the innovation form (for a single consensus step), as (also showed as a result as Eqns. (40-41) in the main paper)

$$\hat{\mathbf{x}}_{i}^{+} = \hat{\mathbf{x}}_{i}^{-} + \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i}) \right)^{-1} \left(\mathcal{A}(\mathbf{u}_{i}) - \mathcal{A}(\mathbf{U}_{i})\hat{\mathbf{x}}_{i}^{-} + \epsilon \sum_{i' \in \mathcal{N}_{c}} \frac{\mathbf{J}_{i'}^{-}}{N_{C}} \left(\hat{\mathbf{x}}_{i'}^{-} - \hat{\mathbf{x}}_{i}^{-} \right) \right)$$

$$(27)$$

$$\mathbf{J}_{i}^{+} = N_{C} \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i}) \right)$$
 (28)

3.2. MTIC

MTIC estimate equations (Eqns. (23-24) in the main paper) for a single consensus step, can be written with the $\mathcal{A}()$ notation as,

$$\hat{\mathbf{x}}_{i}^{+} = \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i}) \right)^{-1} \left(\mathcal{A}(\frac{\mathbf{J}_{i}}{N_{C}}\hat{\mathbf{x}}_{i}^{-}) + \mathcal{A}(\beta_{0}\mathbf{U}_{i}\hat{\mathbf{x}}_{i}^{-}) + \mathcal{A}(\mathbf{u}_{i}) \right)$$
(29)

$$\mathbf{J}_{i}^{+} = N_{C} \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{G}_{i}) \right)$$
(30)

By adding and subtracting $\hat{\mathbf{x}}_i^-$ in RHS of (29) and following the exact same steps of the proof in the previous section we can write the MTIC estimation equations for a single consensus step (also showed as a result as Eqns. (42-43) in the main paper),

$$\hat{\mathbf{x}}_{i}^{+} = \hat{\mathbf{x}}_{i}^{-} + \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{U}_{i})\right)^{-1} \left(\mathcal{A}(\mathbf{u}_{i}) - \mathcal{A}(\mathbf{U}_{i})\hat{\mathbf{x}}_{i}^{-} + \epsilon \sum_{i' \in \mathcal{N}_{i}} \frac{\mathbf{J}_{i'}^{-}}{N_{C}} \left(\hat{\mathbf{x}}_{i'}^{-} - \hat{\mathbf{x}}_{i}^{-}\right) + \mathcal{A}(\beta_{i0}\mathbf{U}_{i}\hat{\mathbf{x}}_{i}^{-})\right)$$
(31)

$$\mathbf{J}_{i}^{+} = N_{C} \left(\mathcal{A}(\frac{\mathbf{J}_{i}^{-}}{N_{C}}) + \mathcal{A}(\mathbf{G}_{i}) \right)$$
(32)