

Appendix

Talha Bozkus and Urbashi Mitra

A. Probability of misdetection for $N_T > 2$ agents

We employ a similar model to 2 agent case:

$$I_{i,t} = I_{i,t,\text{true}} + n_{i,t}, \quad (1)$$

where $n_{i,t} \sim N(0, \sigma_u^2)$ if the state is U, and $n_{i,t} \sim N(0, \sigma_c^2)$ if the state is C, and $\sigma_u > \sigma_c$, where $I_{i,t}$ is the noisy ARSS measurement of agent i due to all other agents. The joint state is C if there is at least one agent i such that $I_{i,t} > I_{thr}$. We define the probability of error as follows:

$$P_{err} = P(\text{detect uncoor} \mid \text{actual coor})P(\text{actual coor}) + P(\text{detect coor} \mid \text{actual uncoor})P(\text{actual uncoor}). \quad (2)$$

$$= P(\text{detect uncoor} \mid \text{actual coor})P(\text{actual coor}) + (1 - P(\text{detect uncoor} \mid \text{actual uncoor}))P(\text{actual uncoor}), \quad (3)$$

where $P(\text{actual coor}) = (1 - \alpha) = \frac{|\mathcal{S}_c|}{|\mathcal{S}|}$ and $P(\text{actual uncoor}) = \alpha = \frac{|\mathcal{S}_u|}{|\mathcal{S}|}$.

We first want to find a lower bound on P_{err} . To do so, we first find a lower bound on $P(\text{detect uncoor} \mid \text{actual coor})$:

$$P(\text{detect uncoor} \mid \text{actual coor}) = P(\max_i \{I_{i,t}\} < I_{thr} \mid \text{actual coor}). \quad (4)$$

$$\geq 1 - \sum_{i=1}^{N_T} P(I_{i,t} > I_{thr} \mid \text{actual coor}). \quad (5)$$

$$= 1 - \sum_{i=1}^{N_T} Q\left(\frac{I_{thr} - I_{i,t,\text{true}}}{\sigma_c}\right) \quad (6)$$

where (4) follows from the criteria for detecting coordination, (5) follows from Boole's inequality and (6) follows from (1). We next find an upper bound on $P(\text{detect uncoor} \mid \text{actual uncoor})$:

$$P(\text{detect uncoor} \mid \text{actual uncoor}) = P(\max_i \{I_{i,t}\} < I_{thr} \mid \text{actual uncoor}). \quad (7)$$

$$= P(I_{i,1} < I_{thr}, I_{i,2} < I_{thr}, \dots \mid \text{actual uncoor}). \quad (8)$$

$$\leq \min_i P(I_{i,i} < I_{thr} \mid \text{actual uncoor}). \quad (9)$$

$$= 1 - \min_i Q\left(\frac{I_{thr} - I_{i,t,\text{true}}}{\sigma_u}\right). \quad (10)$$

$$= 1 - Q\left(\frac{I_{thr} - I_{i',t,\text{true}}}{\sigma_u}\right), \quad (11)$$

where (7) follows from the criteria for detecting coordination, (8) and (9) follow from the properties of probability, (10) follows from (1), (11) follows from the notation that $I_{i',t,\text{true}} = \min_{i=1,\dots,N} I_{i,t,\text{true}}$. By combining (6) and (11), we obtain:

$$P_{err} = f(I_{thr}) \geq \left(1 - \sum_{i=1}^{N_T} Q\left(\frac{I_{thr} - I_{i,t,\text{true}}}{\sigma_c}\right)\right)(1 - \alpha) + Q\left(\frac{I_{thr} - I_{i',t,\text{true}}}{\sigma_u}\right)\alpha, \quad (12)$$

We use the second-order Taylor expansion of the Q -function around $x = 0$:

$$Q(x) = \frac{1}{2} - \frac{x}{\sqrt{2\pi}} - \frac{x^3}{6\sqrt{2\pi}}. \quad (13)$$

Combining (13) with 12, we obtain:

$$f(I_{thr}) \geq \left(1 - \sum_{i=1}^{N_T} \left(\frac{1}{2} - \frac{I_{thr} - I_{i,t,\text{true}}}{\sqrt{2\pi}\sigma_c} - \frac{(I_{thr} - I_{i,t,\text{true}})^3}{6\sqrt{2\pi}\sigma_c^3}\right)\right)(1 - \alpha) + \left(\frac{1}{2} - \frac{I_{thr} - I_{i',t,\text{true}}}{\sqrt{2\pi}\sigma_u} - \frac{(I_{thr} - I_{i',t,\text{true}})^3}{6\sqrt{2\pi}\sigma_u^3}\right)\alpha. \quad (14)$$

Lets call $I_{i',t,true} = c$. To find the value of I_{thr} that minimizes this term, we take the gradient of RHS of (14) with respect to I_{thr} :

$$\frac{d}{dI_{thr}} f(I_{thr}) = (1 - \alpha) \sum_{i=1}^{N_T} \left(\frac{1}{\sqrt{2\pi}\sigma_c} + \frac{(I_{thr} - I_{i,t,true})^2}{2\sqrt{2\pi}\sigma_c^3} \right) + \alpha \left(-\frac{1}{\sqrt{2\pi}\sigma_u} - \frac{(I_{thr} - c)^2}{2\sqrt{2\pi}\sigma_u^3} \right). \quad (15)$$

$$= \frac{2(1 - \alpha)N}{\sigma_c} + \frac{(1 - \alpha)}{\sigma_c^3} \sum_{i=1}^{N_T} (I_{thr} - I_{i,t,true})^2 - \frac{2\alpha}{\sigma_u} - \frac{\alpha}{\sigma_u^3} (I_{thr} - c)^2. \quad (16)$$

$$= \frac{2(1 - \alpha)N}{\sigma_c} - \frac{2\alpha}{\sigma_u} + \frac{(1 - \alpha)N}{\sigma_c^3} I_{thr}^2 - \frac{2(1 - \alpha)}{\sigma_c^3} \sum_{i=1}^{N_T} I_{i,t,true} I_{thr} + \frac{(1 - \alpha)}{\sigma_c^3} \sum_{i=1}^{N_T} I_{i,t,true}^2 - \frac{\alpha}{\sigma_u^3} I_{thr}^2 + \frac{2\alpha}{\sigma_u^3} I_{thr} c - \frac{\alpha}{\sigma_u^3} c^2. \quad (17)$$

$$= AI_{thr}^2 + BI_{thr} + C, \quad (18)$$

where

$$A = \frac{(1 - \alpha)N}{\sigma_c^3} - \frac{\alpha}{\sigma_u^3}. \quad (19)$$

$$B = -\frac{2(1 - \alpha)}{\sigma_c^3} \sum_{i=1}^{N_T} I_{i,t,true} + \frac{2\alpha}{\sigma_u^3} c. \quad (20)$$

$$C = \frac{2(1 - \alpha)N}{\sigma_c} - \frac{2\alpha}{\sigma_u} + \frac{(1 - \alpha)}{\sigma_c^3} \sum_{i=1}^{N_T} I_{i,t,true}^2 - \frac{\alpha}{\sigma_u^3} c^2. \quad (21)$$

Herein, the critical points are $I_{thr} = \frac{-B \pm \sqrt{B^2 - 4AC}}{2A}$. One can show that $I_{thr}^* = \frac{-B - \sqrt{B^2 - 4AC}}{2A}$ is a global minimizer as $\frac{d^2}{dI_{thr}^2} |_{I_{thr}=I_{thr}^*} > 0$, which follows from $|\mathcal{S}_C| \sigma_u > |\mathcal{S}_U| \sigma_c$, $c < \sum_{i=1}^{N_T} I_{i,t,true}^2$ and $N > 1$. Thus, the lower bound on P_{err} is given by:

$$P_{err} \geq \left(1 - \sum_{i=1}^{N_T} Q\left(\frac{I_{thr}^* - I_{i,t,true}}{\sigma_c}\right) \right) \frac{|\mathcal{S}_c|}{|\mathcal{S}|} + Q\left(\frac{I_{thr}^* - I_{i',t,true}}{\sigma_u}\right) \frac{|\mathcal{S}_u|}{|\mathcal{S}|}. \quad (22)$$

1) *Upper bound:* We take a very similar approach to the lower bound. We first find a lower bound on $P(\text{detect uncoor} | \text{actual uncoor})$:

$$P(\text{detect uncoor} | \text{actual uncoor}) \geq 1 - \sum_{i=1}^{N_T} Q\left(\frac{I_{thr} - I_{i,t,true}}{\sigma_u}\right), \quad (23)$$

which follows from (6). We next find the upper bound on $P(\text{detect uncoor} | \text{actual coor})$:

$$P(\text{detect uncoor} | \text{actual coor}) \leq 1 - Q\left(\frac{I_{thr} - I_{i',t,true}}{\sigma_c}\right), \quad (24)$$

which follows from (11). Combining (23) with (24), we obtain:

$$P_{err} = f(I_{thr}) \leq \sum_{i=1}^{N_T} Q\left(\frac{I_{thr} - I_{i,t,true}}{\sigma_u}\right) \alpha + \left(1 - Q\left(\frac{I_{thr} - I_{i',t,true}}{\sigma_c}\right) \right) (1 - \alpha). \quad (25)$$

If we use (13) in (25), and take the gradient with respect to I_{thr} , we obtain:

$$\frac{d}{dI_{thr}} f(I_{thr}) = \tilde{A} I_{thr}^2 + \tilde{B} I_{thr} + \tilde{C}, \quad (26)$$

where

$$\tilde{A} = \frac{-\alpha N}{\sigma_u^3} + \frac{(1 - \alpha)}{\sigma_c^3} \quad (27)$$

$$\tilde{B} = \frac{2\alpha}{\sigma_u^3} \sum_{i=1}^{N_T} I_{i,t,true} - \frac{2(1 - \alpha)}{\sigma_c^3} c \quad (28)$$

$$\tilde{C} = \frac{-2\alpha N}{\sigma_u} + \frac{2(1 - \alpha)}{\sigma_c} - \frac{\alpha}{\sigma_u^3} \sum_{i=1}^{N_T} I_{i,t,true}^2 + \frac{(1 - \alpha)}{\sigma_c^3} c^2 \quad (29)$$

One can show that $I_{thr}^{**} = \frac{-\tilde{B} + \sqrt{\tilde{B}^2 - 4\tilde{A}\tilde{C}}}{2\tilde{A}}$ is a global maximizer as $\frac{d^2}{dI_{thr}^2}|_{I_{thr}=I_{thr}^{**}} < 0$. Thus, the upper bound on P_{err} is given by:

$$P_{err} \leq \sum_{i=1}^{N_T} Q\left(\frac{I_{thr}^{**} - I_{i,t,true}}{\sigma_u}\right) \frac{|S_u|}{|S|} + \left(1 - Q\left(\frac{I_{thr}^{**} - I_{i',t,true}}{\sigma_c}\right)\right) \frac{|S_c|}{|S|}. \quad (30)$$

Combining (22) and (30), we can bound P_{err} for arbitrary N agents:

$$P_{err} \in \left[\left(1 - \sum_{i=1}^{N_T} Q\left(\frac{I_{thr}^* - I_{i,t,true}}{\sigma_c}\right)\right) \frac{|S_c|}{|S|} + Q\left(\frac{I_{thr}^* - I_{i',t,true}}{\sigma_u}\right) \frac{|S_u|}{|S|}, \right. \quad (31)$$

$$\left. \sum_{i=1}^{N_T} Q\left(\frac{I_{thr}^{**} - I_{i,t,true}}{\sigma_u}\right) \frac{|S_u|}{|S|} + \left(1 - Q\left(\frac{I_{thr}^{**} - I_{i',t,true}}{\sigma_c}\right)\right) \frac{|S_c|}{|S|} \right], \quad (32)$$

where

$$I_{thr}^* = \frac{-B - \sqrt{B^2 - 4AC}}{2A}. \quad (33)$$

$$I_{thr}^{**} = \frac{-\tilde{B} + \sqrt{\tilde{B}^2 - 4\tilde{A}\tilde{C}}}{2\tilde{A}}. \quad (34)$$

$$A = \frac{(1-\alpha)N}{\sigma_c^3} - \frac{\alpha}{\sigma_u^3}. \quad (35)$$

$$B = -\frac{2(1-\alpha)}{\sigma_c^3} \sum_{i=1}^{N_T} I_{i,t,true} + \frac{2\alpha}{\sigma_u^3} I_{i',t,true}. \quad (36)$$

$$C = \frac{2(1-\alpha)N}{\sigma_c} - \frac{2\alpha}{\sigma_u} + \frac{(1-\alpha)}{\sigma_c^3} \sum_{i=1}^{N_T} I_{i,t,true}^2 - \frac{\alpha}{\sigma_u^3} I_{i',t,true}^2. \quad (37)$$

$$\tilde{A} = \frac{-\alpha N}{\sigma_u^3} + \frac{(1-\alpha)}{\sigma_c^3}. \quad (38)$$

$$\tilde{B} = \frac{2\alpha}{\sigma_u^3} \sum_{i=1}^{N_T} I_{i,t,true} - \frac{2(1-\alpha)}{\sigma_c^3} I_{i',t,true}. \quad (39)$$

$$\tilde{C} = \frac{-2\alpha N}{\sigma_u} + \frac{2(1-\alpha)}{\sigma_c} - \frac{\alpha}{\sigma_u^3} \sum_{i=1}^{N_T} I_{i,t,true}^2 + \frac{(1-\alpha)}{\sigma_c^3} I_{i',t,true}^2. \quad (40)$$

$$I_{i',t,true} = \min_{i=1,\dots,N} I_{i,t,true}. \quad (41)$$