

APPENDIX A  
PROOF OF LEMMA 1

According to the definition of  $Q'_a(t) = \max\{Q_m(t), m \in \mathcal{M}_a\}$ , it is can be derived that

$$\Pr(Q_m(t) \leq Q'_a(t)) = 1, \forall m \in \mathcal{M}_a, t, \quad (24)$$

which can also be denoted by  $Q_m(t) \leq Q'_a(t), \forall m \in \mathcal{M}_a, t$ . We take expectations of both sides. According to the monotonicity of expectation, there is

$$\mathbb{E}\{Q_m(t)\} \leq \mathbb{E}\{Q'_a(t)\}, \forall m \in \mathcal{M}_a, t. \quad (25)$$

Similarly, the inequality can also be extended to

$$\lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^T \mathbb{E}\{Q_m(t)\} \leq \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^{T-1} \mathbb{E}\{Q'_a(t)\}, \forall m \in \mathcal{M}_a. \quad (26)$$

Thus, if the constraint  $\lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^{T-1} \mathbb{E}\{Q'_a(t)\} < \infty, \forall a \in \mathcal{A}$  is satisfied, there must be

$$\lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^{T-1} \mathbb{E}\{Q_m(t)\} \leq \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^{T-1} \mathbb{E}\{Q'_a(t)\} \leq \infty, \quad (27)$$

This proved that if the constraint  $\overline{Q'_a(t)} \triangleq \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^{T-1} \mathbb{E}\{Q'_a(t)\} < \infty, \forall a \in \mathcal{A}$  is satisfied, the constraint  $\overline{Q_m(t)} \triangleq \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^{T-1} \mathbb{E}\{Q_m(t)\} < \infty, \forall m \in \mathcal{M}$  satisfies.

This completes the proof.

APPENDIX B  
PROOF OF LEMMA 2

Recalling the equation (10), we can denote the  $\Delta(\Theta(t))$  by

$$\begin{aligned} \Delta(\Theta(t)) &= \mathbb{E} \left\{ \frac{1}{2} \sum_{a=1}^A Q'_a(t+1)^2 - \frac{1}{2} \sum_{a=1}^A Q'_a(t)^2 \right\} \\ &= \mathbb{E} \left\{ \frac{1}{2} \sum_{a=1}^A (y'_a(t) + Q'_a(t))^2 - \frac{1}{2} \sum_{a=1}^A Q'_a(t)^2 \right\} \\ &= \mathbb{E} \left\{ \frac{1}{2} \sum_{a=1}^A (y'_a(t))^2 + \sum_{a=1}^A Q'_a(t) y'_a(t) \right\}. \end{aligned} \quad (28)$$

According to the definition of  $y'_a(t)$  in (10), we have

$$\begin{aligned} y'_a(t) &= Q'_a(t+1) - Q'_a(t) \\ &= \max_{m \in \mathcal{M}_a} Q_m(t+1) - \max_{m \in \mathcal{M}_a} Q_m(t) \end{aligned} \quad (29)$$

According to (6), considering that each  $m \in m \in \mathcal{M}_a$ , there is  $y_m(t) \leq \max_{m \in \mathcal{M}_a} y_m(t)$ , and

$$Q_m(t+1) = Q_m(t) + y_m(t) \leq Q_m(t) + \max_{m \in \mathcal{M}_a} y_m(t), \forall m \in \mathcal{M}_a. \quad (30)$$

Based on (30), we have

$$\begin{aligned} \max_{m \in \mathcal{M}_a} Q_m(t+1) &\leq \max_{m \in \mathcal{M}_a} \left( Q_m(t) + \max_{m \in \mathcal{M}_a} y_m(t) \right) \\ &\leq \max_{m \in \mathcal{M}_a} Q_m(t) + \max_{m \in \mathcal{M}_a} y_m(t). \end{aligned} \quad (31)$$

Recalling (29), there is

$$\begin{aligned} y'_a(t) &= \max_{m \in \mathcal{M}_a} Q_m(t+1) - \max_{m \in \mathcal{M}_a} Q_m(t) \\ &\leq \max_{m \in \mathcal{M}_a} y_m(t). \end{aligned} \quad (32)$$

According to (5), considering  $x_m(t) \leq Q_m(t)$  in (7, there is  $Q_m(t) + A_m(t) - x_m(t) \geq 0$  and  $Q_m(t+1) = Q_m(t) + A_m(t) - x_m(t)$ . Thus, there is

$$y_m(t) = Q_m(t+1) - Q_m(t) = A_m(t) - x_m(t). \quad (33)$$

Associated with (2) and (7), we have

$$\begin{aligned} y_m(t) &= Q_m(t+1) - Q_m(t) = A_m(t) - x_m(t) \\ &\leq D_m, \forall m \in \mathcal{M}. \end{aligned} \quad (34)$$

Therefore, the (32) can be rewritten as

$$y'_a(t) \leq \max_{m \in \mathcal{M}_a} y_m(t) \leq D_m = D^a, \forall m \in \mathcal{M}_a. \quad (35)$$

and also

$$\frac{1}{2} \sum_{a=1}^A (y'_a(t))^2 \leq \frac{1}{2} \sum_{a=1}^A (D^a)^2 = B. \quad (36)$$

Back to (28), we have

$$\Delta(\Theta(t)) \leq \mathbb{E} \left\{ B + \sum_{a=1}^A Q'_a(t) y'_a(t) | \Theta(t) \right\}. \quad (37)$$

and

$$\Delta_V(\Theta(t)) \leq B + \mathbb{E} \left\{ \sum_{a=1}^A Q'_a(t) y'_a(t) | \Theta(t) \right\} + V \cdot \mathbb{E}\{P(t) | \Theta(t)\}, \quad (38)$$

This completes the proof.

APPENDIX C  
PROOF OF LEMMA 3

Due to we have obtained  $\Theta(t), \{Q_m(t)\}, \{A_m(t)\}, \mathbf{b}(t)$  in each slot before we solve the problem, the constraints (4), (7) are both linear constraints. And for the objective  $\sum_{a=1}^A Q'_a(t) y'_a(t)$ , we have:

$$\begin{aligned} \mathbb{E} \left\{ \sum_{a=1}^A Q'_a(t) y'_a(t) \right\} &= \mathbb{E} \left\{ \sum_{a=1}^A Q'_a(t) (Q'_a(t+1) - Q'_a(t)) \right\} \\ &= \mathbb{E} \left\{ \sum_{a=1}^A Q'_a(t) \left( \max_{m \in \mathcal{M}_a} Q_m(t+1) - \max_{m \in \mathcal{M}_a} Q_m(t) \right) \right\} \\ &= \sum_{a=1}^A Q'_a(t) \left( \max_{m \in \mathcal{M}_a} \mathbb{E}\{Q_m(t+1)\} - \max_{m \in \mathcal{M}_a} Q_m(t) \right) \\ &= \sum_{a=1}^A Q'_a(t) \left( \max_{m \in \mathcal{M}_a} (Q_m(t) + A_m(t) - x_m(t)) - \max_{m \in \mathcal{M}_a} Q_m(t) \right) \\ &= \sum_{a=1}^A Q'_a(t) \max_{m \in \mathcal{M}_a} (-x_m(t)) + C. \end{aligned} \quad (39)$$

Considering the  $\max_{m \in \mathcal{M}_a} (-x_m(t))$  is convex, the weighted sum of these convex functions is also convex. Accompanied with the linear constraints, we can prove this problem is convex. This completes the proof.

APPENDIX D  
PROOF OF THEOREM 1