

This is the online technical report of the manuscript “Achieving Asymptotic Efficiency via Instantaneous Weighted Alpha-Fairness in Online Allocation” submitted to IEEE/ACM ToN. Appendix A provides the proof of Lemma 2 and Lemma 3. Appendix B then provides the proof of Theorem 1 based on the two lemmas.

APPENDIX A

Proof of Lemma 2. Note that $\mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu})$ is δ_t -strongly concave in $\boldsymbol{\mu}$. This implies that

$$\mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}) - \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \leq \langle \boldsymbol{\mu} - \boldsymbol{\mu}_t, \nabla_{\boldsymbol{\mu}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \rangle - \frac{\delta_t}{2} \|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2. \quad (29)$$

Based on the updating rule (16) for $\boldsymbol{\mu}_t$, we have

$$\begin{aligned} \|\boldsymbol{\mu} - \boldsymbol{\mu}_{t+1}\|_2^2 &\leq \|\boldsymbol{\mu} - \boldsymbol{\mu}_t - \beta_t \nabla_{\boldsymbol{\mu}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t)\|_2^2 \\ &= \|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2 + \beta_t^2 \|\nabla_{\boldsymbol{\mu}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t)\|_2^2 - 2\beta_t \langle \boldsymbol{\mu} - \boldsymbol{\mu}_t, \nabla_{\boldsymbol{\mu}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \rangle, \end{aligned} \quad (30)$$

which leads to

$$\langle \boldsymbol{\mu} - \boldsymbol{\mu}_t, \nabla_{\boldsymbol{\mu}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \rangle \leq \frac{\|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2 - \|\boldsymbol{\mu} - \boldsymbol{\mu}_{t+1}\|_2^2}{2\beta_t} + \frac{\beta_t \|\nabla_{\boldsymbol{\mu}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t)\|_2^2}{2}. \quad (31)$$

Substituting (31) into (29) yields

$$\mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}) - \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \leq \frac{\|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2 - \|\boldsymbol{\mu} - \boldsymbol{\mu}_{t+1}\|_2^2}{2\beta_t} + \frac{\beta_t \|\nabla_{\boldsymbol{\mu}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t)\|_2^2}{2} - \frac{\delta_t}{2} \|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2. \quad (32)$$

Summing over $t \in \{1, 2, \dots, T\}$ leads to

$$\sum_{t=1}^T \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}) - \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \leq \underbrace{\sum_{t=1}^T \frac{\beta_t \|\nabla_{\boldsymbol{\mu}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t)\|_2^2}{2}}_{\text{Part I}} + \underbrace{\sum_{t=1}^T \left(\frac{\|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2 - \|\boldsymbol{\mu} - \boldsymbol{\mu}_{t+1}\|_2^2}{2\beta_t} - \frac{\delta_t}{2} \|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2 \right)}_{\text{Part II}}. \quad (33)$$

Next we upper bound the right-hand-side (RHS) of (33) in two steps.

- First, we derive an upper bound for Part I of the RHS of (33) as follows:

$$\sum_{t=1}^T \frac{\beta_t \|\nabla_{\boldsymbol{\mu}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t)\|_2^2}{2} \leq \sum_{t=1}^T \beta_t (MNA^2G^2 + \delta_t^2 \|\boldsymbol{\mu}_t\|_2^2) = MNA^2G^2 J_T(\boldsymbol{\beta}) + \sum_{t=1}^T \beta_t \delta_t^2 \|\boldsymbol{\mu}_t\|_2^2, \quad (34)$$

where the inequality follows from $(a_1 + a_2 + \dots + a_n)^2 \leq n(a_1^2 + a_2^2 + \dots + a_n^2)$.

- Second, we derive an upper bound for Part II of the RHS of (33) as follows:

$$\begin{aligned} &\sum_{t=1}^T \left(\frac{\|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2 - \|\boldsymbol{\mu} - \boldsymbol{\mu}_{t+1}\|_2^2}{2\beta_t} - \frac{\delta_t}{2} \|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2 \right) \\ &= \frac{\|\boldsymbol{\mu} - \boldsymbol{\mu}_1\|_2^2}{2\beta_1} - \frac{\|\boldsymbol{\mu} - \boldsymbol{\mu}_{T+1}\|_2^2}{2\beta_T} + \sum_{t=2}^T \frac{\|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2}{2} \left(\frac{1}{\beta_t} - \frac{1}{\beta_{t-1}} \right) - \sum_{t=1}^T \frac{\delta_t}{2} \|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2 \\ &\stackrel{(a)}{\leq} \frac{\|\boldsymbol{\mu}\|_2^2}{2} \left(\frac{1}{\beta_1} - \delta_1 \right) + \sum_{t=2}^T \frac{\|\boldsymbol{\mu} - \boldsymbol{\mu}_t\|_2^2}{2} \left(\frac{1}{\beta_t} - \frac{1}{\beta_{t-1}} - \delta_t \right) \stackrel{(b)}{\leq} \frac{\|\boldsymbol{\mu}\|_2^2}{2} \left(\frac{1}{\beta_1} - \delta_1 \right) \end{aligned} \quad (35)$$

where the inequality (a) follows from $\frac{\|\boldsymbol{\mu} - \boldsymbol{\mu}_{T+1}\|_2^2}{2\beta_T} \geq 0$ and $\boldsymbol{\mu}_1 = \mathbf{0}_M$. The inequality (b) follows from Lemma 4 (presented in Appendix B).

Substituting (34) and (35) into (33) completes the proof. \square

Proof of Lemma 3. Note that $\mathcal{L}_t(\cdot; \boldsymbol{\mu}_t)$ is convex in \mathbf{x} . This implies that

$$\sum_{t=1}^T \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) - \mathcal{L}_t(\mathbf{x}; \boldsymbol{\mu}_t) \leq \sum_{t=1}^T \langle \hat{\mathbf{x}}_t - \mathbf{x}, \nabla_{\mathbf{x}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \rangle. \quad (36)$$

Next we will derive an upper bound for the RHS in (36) based on the potential function $\Phi_t(\alpha)$, i.e.,

$$\Phi_t(\alpha) \triangleq \alpha \ln \left(\frac{1}{N} \sum_{n=1}^N \exp \left(-\frac{1}{\alpha} \sum_{s=1}^t \frac{\partial \mathcal{L}_s(\hat{\mathbf{x}}_s; \boldsymbol{\mu}_s)}{\partial x_n} \right) \right). \quad (37)$$

For any $t > 1$, we have

$$\begin{aligned}
\Phi_t(\alpha_t) - \Phi_{t-1}(\alpha_t) &= \alpha_t \ln \left(\frac{\sum_{n=1}^N \exp \left(-\frac{1}{\alpha_t} \sum_{s=1}^t \frac{\partial \mathcal{L}_s(\hat{\mathbf{x}}_s; \boldsymbol{\mu}_s)}{\partial x_n} \right)}{\sum_{n=1}^N \exp \left(-\frac{1}{\alpha_t} \sum_{s=1}^{t-1} \frac{\partial \mathcal{L}_s(\hat{\mathbf{x}}_s; \boldsymbol{\mu}_s)}{\partial x_n} \right)} \right) \\
&\stackrel{(a)}{=} \alpha_t \ln \left(\frac{\sum_{n=1}^N w_{t,n}^{1/\alpha_t} \cdot \exp \left(-\frac{1}{\alpha_t} \frac{\partial \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t)}{\partial x_n} \right)}{\sum_{i=1}^N w_{t,i}^{1/\alpha_t}} \right) \\
&\stackrel{(b)}{=} \alpha_t \ln \left(\sum_{n=1}^N \frac{\hat{x}_{t,n}}{A} \exp \left(-\frac{1}{\alpha_t} \frac{\partial \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t)}{\partial x_n} \right) \right) \\
&\stackrel{(c)}{\leq} -\frac{\langle \hat{\mathbf{x}}_t, \nabla_{\mathbf{x}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \rangle}{A} + \frac{(1 + \|\boldsymbol{\mu}_t\|_1)^2 G^2}{2\alpha_t} \\
&\stackrel{(d)}{\leq} -\frac{\langle \hat{\mathbf{x}}_t, \nabla_{\mathbf{x}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \rangle}{A} + \frac{(1 + M)(1 + \|\boldsymbol{\mu}_t\|_2^2) G^2}{2\alpha_t},
\end{aligned} \tag{38}$$

where the equality (a) follows from the weight vector's updating rule (15), and the equality (b) follows from (12). The inequality (c) follows from Hoeffding's inequality and $\|[\nabla_{\mathbf{x}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t)]_{(n)}\| \leq (1 + \|\boldsymbol{\mu}_t\|_1)G$. The inequality (d) follows from $(a_1 + a_2 + \dots + a_k)^2 \leq k(a_1^2 + a_2^2 + \dots + a_k^2)$.

Rearranging the terms in (38) yields

$$\frac{\langle \hat{\mathbf{x}}_t, \nabla_{\mathbf{x}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \rangle}{A} \leq \frac{(1 + M)(1 + \|\boldsymbol{\mu}_t\|_2^2) G^2}{2\alpha_t} + \Phi_{t-1}(\alpha_t) - \Phi_t(\alpha_t). \tag{39}$$

Summing over $t \in \{1, 2, \dots, T\}$ leads to

$$\begin{aligned}
\sum_{t=1}^T \frac{\langle \hat{\mathbf{x}}_t, \nabla_{\mathbf{x}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \rangle}{A} &\leq \sum_{t=1}^T [\Phi_{t-1}(\alpha_t) - \Phi_t(\alpha_t)] + \sum_{t=1}^T \frac{(1 + M)(1 + \|\boldsymbol{\mu}_t\|_2^2) G^2}{2\alpha_t} \\
&\stackrel{(a)}{=} \Phi_0(\alpha_1) - \Phi_T(\alpha_T) + \sum_{t=1}^{T-1} [\Phi_t(\alpha_{t+1}) - \Phi_t(\alpha_t)] + \sum_{t=1}^T \frac{(1 + M)(1 + \|\boldsymbol{\mu}_t\|_2^2) G^2}{2\alpha_t} \\
&\stackrel{(b)}{\leq} \Phi_0(\alpha_1) - \Phi_T(\alpha_T) + \sum_{t=1}^T \frac{(1 + M)(1 + \|\boldsymbol{\mu}_t\|_2^2) G^2}{2\alpha_t} \\
&\stackrel{(c)}{=} -\Phi_T(\alpha_T) + \sum_{t=1}^T \frac{(1 + M)(1 + \|\boldsymbol{\mu}_t\|_2^2) G^2}{2\alpha_t},
\end{aligned} \tag{40}$$

where the equality (a) follows from Abel transformation. The inequality (b) follows from that $\Phi_t(\alpha)$ is decreasing in α and the sequence $\{\alpha_t : \forall t \in \mathcal{T}\}$ is increasing in t according to (14). The equality (c) follows from $\Phi_0(\alpha_1) = 0$.

Next we derive a lower bound for $\Phi_T(\alpha_T)$. Speciality, we have

$$\begin{aligned}
\Phi_T(\alpha_T) &= \alpha_T \ln \left(\frac{1}{N} \sum_{n=1}^N \exp \left(-\frac{1}{\alpha_T} \sum_{t=1}^T [\nabla_{\mathbf{x}} \mathcal{L}_t(\mathbf{x}_t; \boldsymbol{\mu}_t)]_{(n)} \right) \right) \\
&\stackrel{(a)}{\geq} \alpha_T \ln \left(\sum_{n=1}^N \frac{x_n}{A} \exp \left(-\frac{1}{\alpha_T} \sum_{t=1}^T [\nabla_{\mathbf{x}} \mathcal{L}_t(\mathbf{x}_t; \boldsymbol{\mu}_t)]_{(n)} \right) \right) - \alpha_T \ln(N) \\
&\stackrel{(b)}{\geq} -\sum_{n=1}^N \frac{x_n}{A} \sum_{t=1}^T [\nabla_{\mathbf{x}} \mathcal{L}_t(\mathbf{x}_t; \boldsymbol{\mu}_t)]_{(n)} - \alpha_T \ln(N) \\
&= -\sum_{t=1}^T \frac{\langle \mathbf{x}, \nabla_{\mathbf{x}} \mathcal{L}_t(\mathbf{x}_t; \boldsymbol{\mu}_t) \rangle}{A} - \alpha_T \ln(N)
\end{aligned} \tag{41}$$

where the inequality (a) follows from $\|\mathbf{x}\|_1 = A$ and $\mathbf{x} \geq 0$. The inequality (b) follows from Jensen's inequality.

Combining the above two aspects, we have

$$\sum_{t=1}^T \frac{\langle \hat{\mathbf{x}}_t - \mathbf{x}, \nabla_{\mathbf{x}} \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) \rangle}{A} \leq \alpha_T \ln(N) + \sum_{t=1}^T \frac{(1 + M)(1 + \|\boldsymbol{\mu}_t\|_2^2) G^2}{2\alpha_t}. \tag{42}$$

Substituting (42) into (36) completes the proof. \square

APPENDIX B

Proof of Theorem 1. We derive an lower bound for $\sum_{t=1}^T \mathbb{E} [\mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}) - \mathcal{L}_t(\mathbf{x}; \boldsymbol{\mu}_t)]$ in two steps. We expand $\sum_{t=1}^T \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}) - \mathcal{L}_t(\mathbf{x}; \boldsymbol{\mu}_t)$, and obtain

$$\sum_{t=1}^T \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}) - \mathcal{L}_t(\mathbf{x}; \boldsymbol{\mu}_t) = \sum_{t=1}^T f_t(\hat{\mathbf{x}}_t) - f_t(\mathbf{x}) + \sum_{m=1}^M \left[\mu_m \sum_{t=1}^T g_{t,m}(\hat{\mathbf{x}}_t) - \sum_{t=1}^T \mu_{t,m} g_{t,m}(\mathbf{x}) \right] - \sum_{t=1}^T \frac{\delta_t \|\boldsymbol{\mu}\|_2^2}{2} + \sum_{t=1}^T \frac{\delta_t \|\boldsymbol{\mu}_t\|_2^2}{2}. \quad (43)$$

Taking the expectation on both sides over $\{\theta_t : \forall t \in \mathcal{T}\}$ leads to

$$\begin{aligned} & \sum_{t=1}^T \mathbb{E} [\mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}) - \mathcal{L}_t(\mathbf{x}; \boldsymbol{\mu}_t)] \\ &= \text{Reg}_T(\mathcal{U}) + \sum_{m=1}^M \left[\mu_m \sum_{t=1}^T \mathbb{E}[g_{t,m}(\hat{\mathbf{x}}_t)] - \sum_{t=1}^T \mu_{t,m} \mathbb{E}[g_{t,m}(\mathbf{x})] \right] - \sum_{t=1}^T \frac{\delta_t \|\boldsymbol{\mu}\|_2^2}{2} + \sum_{t=1}^T \frac{\delta_t \mathbb{E} \|\boldsymbol{\mu}_t\|_2^2}{2} \\ &\geq \text{Reg}_T(\mathcal{U}) + \sum_{m=1}^M \mu_m \sum_{t=1}^T \mathbb{E}[g_{t,m}(\hat{\mathbf{x}}_t)] - \sum_{t=1}^T \frac{\delta_t \|\boldsymbol{\mu}\|_2^2}{2} + \sum_{t=1}^T \frac{\delta_t \mathbb{E} \|\boldsymbol{\mu}_t\|_2^2}{2}. \end{aligned} \quad (44)$$

Next we derive an upper bound for $\sum_{t=1}^T [\mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}) - \mathcal{L}_t(\mathbf{x}; \boldsymbol{\mu}_t)]$ based on Lemma 2 and Lemma 3 as follows

$$\begin{aligned} \sum_{t=1}^T \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}) - \mathcal{L}_t(\mathbf{x}; \boldsymbol{\mu}_t) &= \sum_{t=1}^T \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}) - \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) + \sum_{t=1}^T \mathcal{L}_t(\hat{\mathbf{x}}_t; \boldsymbol{\mu}_t) - \mathcal{L}_t(\mathbf{x}; \boldsymbol{\mu}_t) \\ &\leq \frac{\|\boldsymbol{\mu}\|_2^2}{2} \left(\frac{1}{\beta_1} - \delta_1 \right) + M N A^2 G^2 J_T(\boldsymbol{\beta}) + \sum_{t=1}^T \beta_t \delta_t^2 \|\boldsymbol{\mu}_t\|_2^2 \\ &\quad + A \left[\alpha_T \ln(N) + \frac{(1+M)G^2}{2} \left(J_T(\boldsymbol{\alpha}) + \sum_{t=1}^T \frac{\|\boldsymbol{\mu}_t\|_2^2}{\alpha_t} \right) \right]. \end{aligned} \quad (45)$$

Combining (44) and (45) yields

$$\begin{aligned} & \text{Reg}_T(\mathcal{U}) + \sum_{m=1}^M \mu_m \sum_{t=1}^T \mathbb{E}[g_{t,m}(\hat{\mathbf{x}}_t)] - \sum_{t=1}^T \frac{\delta_t \|\boldsymbol{\mu}\|_2^2}{2} + \sum_{t=1}^T \frac{\delta_t \mathbb{E} \|\boldsymbol{\mu}_t\|_2^2}{2} \\ &\leq \frac{\|\boldsymbol{\mu}\|_2^2}{2} \left(\frac{1}{\beta_1} - \delta_1 \right) + M N A^2 G^2 J_T(\boldsymbol{\beta}) + \sum_{t=1}^T \beta_t \delta_t^2 \mathbb{E} \|\boldsymbol{\mu}_t\|_2^2 \\ &\quad + A \alpha_T \ln(N) + \frac{(1+M)G^2 A J_T(\boldsymbol{\alpha})}{2} + \frac{(1+M)G^2 A}{2} \sum_{t=1}^T \frac{\mathbb{E} \|\boldsymbol{\mu}_t\|_2^2}{\alpha_t}. \end{aligned} \quad (46)$$

Rearranging the terms yields

$$\begin{aligned} & \text{Reg}_T(\mathcal{U}) + \sum_{m=1}^M \left[\mu_m \sum_{t=1}^T \mathbb{E}[g_{t,m}(\hat{\mathbf{x}}_t)] - \frac{\frac{1}{\beta_1} - \delta_1 + J_T(\boldsymbol{\delta})}{2} \cdot (\mu_m)^2 \right] \\ &\leq \left[\alpha_T \ln(N) + \frac{(1+M)G^2 J_T(\boldsymbol{\alpha})}{2} \right] A + M N A^2 G^2 J_T(\boldsymbol{\beta}) + \sum_{t=1}^T \mathbb{E} \|\boldsymbol{\mu}_t\|_2^2 \left(\beta_t \delta_t^2 + \frac{(1+M)G^2 A}{2\alpha_t} - \frac{\delta_t}{2} \right) \\ &\leq \left[\alpha_T \ln(N) + \frac{(1+M)G^2 J_T(\boldsymbol{\alpha})}{2} \right] A + M N A^2 G^2 J_T(\boldsymbol{\beta}) \end{aligned} \quad (47)$$

where the last inequality follows from Lemma 4. Maximizing left-hand-side (LHS) of (47) over $\boldsymbol{\mu}$ leads to

$$\text{Reg}_T(\mathcal{U}) + \frac{\sum_{m=1}^M \mathbb{E} \left(\left[\sum_{t=1}^T g_{t,m}(\mathbf{x}_t) \right]^+ \right)^2}{2 \left(\frac{1}{\beta_1} - \delta_1 + J_T(\boldsymbol{\delta}) \right)} \leq \left[\alpha_T \ln(N) + \frac{(1+M)G^2 J_T(\boldsymbol{\alpha})}{2} \right] A + M N A^2 G^2 J_T(\boldsymbol{\beta}). \quad (48)$$

On the LHS of (48), we have

$$\sum_{m=1}^M \mathbb{E} \left(\left[\sum_{t=1}^T g_{t,m}(\mathbf{x}_t) \right]^+ \right)^2 \geq \frac{[\text{Vio}_T(\mathcal{U})]^2}{M}, \quad (49)$$

which follows from the inequality $z_1^2 + z_2^2 + \dots + z_M^2 \geq (z_1 + z_2 + \dots + z_M)^2/M$.

Therefore, we obtain

$$\text{Reg}_T(\mathcal{U}) + \frac{[\text{Vio}_T(\mathcal{U})]^2}{2\left(\frac{1}{\beta_1} - \delta_1 + J_T(\delta)\right)M} \leq \left[\alpha_T \ln(N) + \frac{(1+M)G^2 J_T(\alpha)}{2}\right] A + MNA^2 G^2 J_T(\beta) \quad (50)$$

When $\alpha_t = t^\epsilon G \sqrt{\frac{1+M}{2\ln(N)}}$, we obtain

$$\begin{aligned} \text{Reg}_T(\mathcal{U}) + \frac{[\text{Vio}_T(\mathcal{U})]^2}{2\left(\frac{1}{\beta_1} - \delta_1 + J_T(\delta)\right)M} &\leq \left[\alpha_T \ln(N) + \frac{(1+M)G^2}{2} J_T(\alpha)\right] A + MNA^2 G^2 J_T(\beta) \\ &= \left(T^\epsilon + \frac{T^{1-\epsilon}}{1-\epsilon}\right) AG \sqrt{\frac{(1+M)\ln(N)}{2}} + MNA^2 G^2 J_T(\beta) \\ &\leq \left(T^\epsilon + \frac{T^{1-\epsilon}}{1-\epsilon}\right) AG \sqrt{\frac{(1+M)\ln(N)}{2}} + \frac{MNA G}{3\sqrt{2(1+M)\ln(N)}} \cdot \frac{T^\epsilon}{\epsilon} \\ &= AG \cdot \Omega_T(\epsilon) \end{aligned} \quad (51)$$

Note that $\frac{1}{\beta_1} - \delta_1 = 3AG\sqrt{2(1+M)\ln(N)}$. This leads to

$$\begin{aligned} \text{Vio}_T(\mathcal{U}) &\leq \sqrt{6AGM\sqrt{2(1+M)\ln(N)}\left(1 + \frac{T^{1-\epsilon}}{1-\epsilon}\right)\left(AG \cdot \Omega_T(N, M) + TAG\sqrt{2N}\right)} \\ &\leq AG \sqrt{\frac{24MC}{1-\epsilon} \left[\frac{\Omega_T(\epsilon)}{T} + \sqrt{2N}\right]} \cdot T^{1-\epsilon/2}, \end{aligned} \quad (52)$$

where $C = \sqrt{(1+M)\ln(N)/2}$. This completes the proof of this theorem. \square

Lemma 4. Conditions C1 and C2 hold if the sequence $\{\alpha_t\}_{t \geq 1}$ is increasing, the sequences $\{\delta_t\}_{t \geq 1}$ and $\{\beta_t\}_{t \geq 1}$ satisfy (17).

$$\begin{aligned} \text{C1: } \frac{1}{\beta_t} - \frac{1}{\beta_{t-1}} - \delta_t &\leq 0, \quad \forall t \geq 2 \\ \text{C2: } \beta_t \delta_t^2 + \frac{(1+M)G^2 A}{2\alpha_t} - \frac{\delta_t}{2} &\leq 0, \quad \forall t \geq 2 \end{aligned} \quad (53)$$

Proof of Lemma 4. We prove C1 as follows:

$$\frac{1}{\beta_t} - \frac{1}{\beta_{t-1}} = \delta_t(t+1) - \delta_{t-1}t = (\delta_t - \delta_{t-1})t + \delta_t \leq \delta_t, \quad (54)$$

where the last inequality follows from that the sequence $\{\delta_t\}_{t \geq 1}$ is decreasing (i.e., $\delta_t < \delta_{t-1}$).

We prove C2 as follows:

$$\beta_t \delta_t^2 + \frac{(1+M)G^2 A}{2\alpha_t} - \frac{\delta_t}{2} \stackrel{(a)}{=} \delta_t \left(\frac{1}{t+1} - \frac{1}{2} \right) + \frac{(1+M)G^2 A}{2\alpha_t} \stackrel{(b)}{\leq} -\frac{\delta_t}{6} + \frac{(1+M)G^2 A}{2\alpha_t} \stackrel{(c)}{=} 0, \quad (55)$$

where the equality (a) follows from $\beta_t = \frac{1}{\delta_t(t+1)}$, the inequality (b) follows $t \geq 2$ in C2, the equality (c) follows from $\delta_t = \frac{3(1+M)G^2 A}{\alpha_t}$. This completes the proof. \square