

# Supplementary Materials of TAoI for Remote Inference with Hybrid Language Models

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## I. LEMMAS AND THEOREM

**Lemma 1.** *The value function  $V(\Delta)$  is non-decreasing with  $\Delta$ .*

*Proof:* See Section II-A. □

**Lemma 2.** *The value function  $V(\Delta)$  is concave in  $\Delta$ .*

*Proof:* See Section II-B. □

Since the value function  $V(\Delta)$  is non-decreasing and concave, its slope is non-increasing and lower bounded. The lower bound of the slope of  $V(\Delta)$  is given by the following lemma. Prior to that, we define an auxiliary variable  $l_{min}$  as follows:

$$l_{min} = \min \left( \frac{T_1^u + T_1^c}{p_s}, \frac{T_2^u + T_1^c}{q_s}, \frac{T_1^u + T_2^c}{p_l}, \frac{T_2^u + T_2^c}{q_l} \right). \quad (1)$$

**Lemma 3.** *For any  $\Delta_1, \Delta_2 \in \mathcal{S}$  with  $\Delta_1 \leq \Delta_2$ , we have  $V(\Delta_2) - V(\Delta_1) \geq \frac{L(\hat{\mathbf{a}})}{\epsilon p}(\Delta_2 - \Delta_1)$ , where  $\hat{\mathbf{a}}$  and  $\hat{p}$  are given by*

$$(\hat{\mathbf{a}}, \hat{p}) = \begin{cases} ((0, 0), p_s), & \text{if } l_{min} = \frac{T_1^u + T_1^c}{p_s} \\ ((1, 0), q_s), & \text{if } l_{min} = \frac{T_2^u + T_1^c}{q_s} \\ ((0, 1), p_l), & \text{if } l_{min} = \frac{T_1^u + T_2^c}{p_l} \\ ((1, 1), q_l), & \text{if } l_{min} = \frac{T_2^u + T_2^c}{q_l} \end{cases}. \quad (2)$$

*Proof:* See Section II-C. □

Based on Lemmas 1-3, we can derive the structure of the optimal control policy as stated in the following theorem.

**Theorem 4.** For any  $\Delta_1, \Delta_2 \in \mathcal{S}$  with  $\Delta_1 \leq \Delta_2$ , there exists a stationary deterministic optimal policy with a threshold-based structure, described as follows:

- When  $l_{min} = \frac{T_1^u + T_1^c}{p_s}$  and  $\pi^*(\Delta_1) = (0, 0)$ ,  $\pi^*(\Delta_2) = (0, 0)$ .
- When  $l_{min} = \frac{T_2^u + T_1^c}{q_s}$  and  $\pi^*(\Delta_1) = (1, 0)$ ,  $\pi^*(\Delta_2) = (1, 0)$ .
- When  $l_{min} = \frac{T_1^u + T_2^c}{p_l}$  and  $\pi^*(\Delta_1) = (0, 1)$ ,  $\pi^*(\Delta_2) = (0, 1)$ .
- When  $l_{min} = \frac{T_2^u + T_2^c}{q_l}$  and  $\pi^*(\Delta_1) = (1, 1)$ ,  $\pi^*(\Delta_2) = (1, 1)$ .

*Proof:* See Section II-D. □

## II. PROOFS OF LEMMAS AND THEOREM

### A. Proof of Lemma 1

Based on the value iteration algorithm (VIA) outlined in [1, Ch. 4.3], we utilize mathematical induction to establish the proof of Lemma 1. Initially, we introduce  $Q_k(s, \mathbf{a})$  and  $V_k(s)$  to represent the state-action value function and the state value function at the  $k$ -th iteration, respectively. Particularly,  $Q_k(s, \mathbf{a})$  is defined as

$$Q_k(s, \mathbf{a}) \triangleq \bar{R}(s, \mathbf{a}) + \sum_{s' \in \mathcal{S}} \bar{p}(s'|s, \mathbf{a}) V_k(s'), \quad \forall s \in \mathcal{S}. \quad (3)$$

For any given state  $s$ , the update to the value function can be executed by

$$V_{k+1}(s) = \min_{\mathbf{a} \in \mathcal{A}} Q_k(s, \mathbf{a}), \quad \forall s \in \mathcal{S}. \quad (4)$$

Regardless of how  $V_0(s)$  is initially set, the sequence  $\{V_k(s)\}$  converges to  $V(s)$  that satisfies the Bellman equation, i.e.,

$$\lim_{k \rightarrow \infty} V_k(s) = V(s), \quad \forall s \in \mathcal{S}. \quad (5)$$

Therefore, the monotonicity of  $V(\Delta)$  is validated by showing that, for any two states  $\Delta_1, \Delta_2 \in \mathcal{S}$ , whenever  $\Delta_1 \leq \Delta_2$ , it follows that

$$V_k(\Delta_1) \leq V_k(\Delta_2), \quad k = 0, 1, \dots. \quad (6)$$

Next, we prove (6) using mathematical induction. Without loss of generality, we set  $V_0(\Delta) = 0$  for each  $\Delta \in \mathcal{S}$ , ensuring that (6) is satisfied at  $k = 0$ . Then, assuming that (6) holds up to  $k > 0$ , we verify whether it holds for  $k + 1$ .

For  $\mathbf{a} = (0, 0)$ , it follows that

$$\begin{aligned} Q_k(\Delta_1, (0, 0)) = & \Delta_1 + \frac{1}{2}(T_1^u + T_1^c - 1) + \frac{\epsilon}{T_1^u + T_1^c} p_s V_k(T_1^u + T_1^c) + \\ & \frac{\epsilon}{T_1^u + T_1^c} (1 - p_s) V_k(\Delta_1 + T_1^u + T_1^c) + \left(1 - \frac{\epsilon}{T_1^u + T_1^c}\right) V_k(\Delta_1), \end{aligned} \quad (7)$$

and

$$\begin{aligned} Q_k(\Delta_2, (0, 0)) = & \Delta_2 + \frac{1}{2}(T_1^u + T_1^c - 1) + \frac{\epsilon}{T_1^u + T_1^c} p_s V_k(T_1^u + T_1^c) + \\ & \frac{\epsilon}{T_1^u + T_1^c} (1 - p_s) V_k(\Delta_2 + T_1^u + T_1^c) + \left(1 - \frac{\epsilon}{T_1^u + T_1^c}\right) V_k(\Delta_2). \end{aligned} \quad (8)$$

Given that  $\Delta_1 + T_1^u + T_1^c \leq \Delta_2 + T_1^u + T_1^c$ ,  $V_k(\Delta_1) \leq V_k(\Delta_2)$  and  $V_k(\Delta_1') \leq V_k(\Delta_2')$ , we can obtain  $Q_k(\Delta_1, (0, 0)) \leq Q_k(\Delta_2, (0, 0))$ .

Similar to  $\mathbf{a} = (0, 0)$ , it can be easily deduced that  $Q_k(\Delta_1, (1, 0)) \leq Q_k(\Delta_2, (1, 0))$ ,  $Q_k(\Delta_1, (0, 1)) \leq Q_k(\Delta_2, (0, 1))$ , and  $Q_k(\Delta_1, (1, 1)) \leq Q_k(\Delta_2, (1, 1))$  according to  $\Delta_1 \leq \Delta_2$  and  $V_k(\Delta_1) \leq V_k(\Delta_2)$ . By (4), we can get that  $V_{k+1}(\Delta_1) \leq V_{k+1}(\Delta_2)$  for any  $k$ .

This concludes the proof of Lemma 1.

## B. Proof of Lemma 2

The concavity of  $V(\Delta)$  with respect to  $\Delta$  can be demonstrated by showing that, for any  $\Delta_1, \Delta_2 \in \mathcal{S}$  and  $w \in N$ , whenever  $\Delta_1 \leq \Delta_2$ , it follows that

$$\begin{aligned} V_k(\Delta_1 + w) - V_k(\Delta_1) &\geq \\ V_k(\Delta_2 + w) - V_k(\Delta_2), k = 0, 1, \dots \end{aligned} \quad (9)$$

Without sacrificing generality, we set  $V_0(\Delta) = 0$  for all  $\Delta \in \mathcal{S}$ , ensuring that (9) is applicable at  $k = 0$ . Then, we assume that (9) holds up till  $k > 0$  and inspect whether it holds for  $k + 1$ . Now, let  $\Delta' = \Delta + w$ ,  $\Delta_1' = \Delta_1 + w$  and  $\Delta_2' = \Delta_2 + w$ . For ease of explanation, we introduce  $\delta Q(\Delta', \Delta, \mathbf{a}) = Q(\Delta', \mathbf{a}) - Q(\Delta, \mathbf{a})$ .

For  $\mathbf{a} = (0, 0)$ , it follows that

$$\begin{aligned} &\delta Q(\Delta_1', \Delta_1, (0, 0)) - \delta Q(\Delta_2', \Delta_2, (0, 0)) \\ &= [\Delta_1 + w + \frac{1}{2}(T_1^u + T_1^c - 1) + \left(1 - \frac{\epsilon}{T_1^u + T_1^c}\right) V_k(\Delta_1 + w) \\ &\quad + \frac{\epsilon}{T_1^u + T_1^c} p_s V_k(T_1^u + T_1^c, I_+) + (1 - p_s) V_k(\Delta_1 + w + T_1^u + T_1^c)] \\ &\quad - [\Delta_1 + \frac{1}{2}(T_1^u + T_1^c - 1) + \left(1 - \frac{\epsilon}{T_1^u + T_1^c}\right) V_k(\Delta_1) \\ &\quad + \frac{\epsilon}{T_1^u + T_1^c} p_s V_k(T_1^u + T_1^c) + (1 - p_s) V_k(\Delta_1 + T_1^u + T_1^c)] \\ &\quad - [\Delta_2 + w + \frac{1}{2}(T_1^u + T_1^c - 1) + \left(1 - \frac{\epsilon}{T_1^u + T_1^c}\right) V_k(\Delta_2 + w) \\ &\quad + \frac{\epsilon}{T_1^u + T_1^c} p_s V_k(T_1^u + T_1^c, I_+) + (1 - p_s) V_k(\Delta_2 + w + T_1^u + T_1^c)] \\ &\quad + [\Delta_2 + \frac{1}{2}(T_1^u + T_1^c - 1) + \left(1 - \frac{\epsilon}{T_1^u + T_1^c}\right) V_k(\Delta_2) \\ &\quad + \frac{\epsilon}{T_1^u + T_1^c} p_s V_k(T_1^u + T_1^c) + (1 - p_s) V_k(\Delta_2 + T_1^u + T_1^c)] \\ &= (1 - \frac{\epsilon}{T_1^u + T_1^c}) [(V_k(\Delta_1 + w) - V_k(\Delta_1)) \\ &\quad - (V_k(\Delta_2 + w) - V_k(\Delta_2))] \\ &\quad + \frac{\epsilon}{T_1^u + T_1^c} (1 - p_0) [(V_k(\Delta_1 + w + T_1^u + T_1^c) - V_k(\Delta_1 + T_1^u + T_1^c)) \\ &\quad - (V_k(\Delta_2 + w + T_1^u + T_1^c) - V_k(\Delta_2 + T_1^u + T_1^c))]. \end{aligned} \quad (10)$$

Given that  $V_k(\Delta_1 + w) - V_k(\Delta_1) \geq V_k(\Delta_2 + w) - V_k(\Delta_2)$  and  $V_k(\Delta_1 + w + 1) - V_k(\Delta_1 + 1) \geq V_k(\Delta_2 + w + 1) - V_k(\Delta_2 + 1)$ , we can easily see that  $\delta Q_k(\Delta'_1, \Delta_1, (0, 0)) - \delta Q_k(\Delta'_2, \Delta_2, (0, 0)) \geq 0$ . Thus,  $Q_k(\Delta, (0, 0))$  is concave in  $\Delta$ .

Similar to  $\mathbf{a} = (0, 0)$ , we can also get that  $Q_k(\Delta, (1, 0))$ ,  $Q_k(\Delta, (0, 1))$ , and  $Q_k(\Delta, (1, 1))$  is concave in  $\Delta$ . Since the value function  $V_{k+1}(\Delta)$  is the minimum of two concave functions, it is also concave in  $\Delta$ . Hence, we have  $V_k(\Delta_1 + w) - V_k(\Delta_1) \geq V_k(\Delta_2 + w) - V_k(\Delta_2)$ , i.e., (9) holds for  $k + 1$ . Therefore, we can show that (9) holds for any  $k$  by induction.

This completes the proof of Lemma 2.

### C. Proof of Lemma 3

The proof follows the same procedure of Lemma 1. The lower bound of  $V(\Delta_2) - V(\Delta_1)$  can be proved by showing that for any  $\Delta_1, \Delta_2 \in \mathcal{S}$ , such that  $\Delta_1 \leq \Delta_2$

$$V_k(\Delta_2) - V_k(\Delta_1) \geq \frac{L(\hat{\mathbf{a}})}{\epsilon p}(\Delta_2 - \Delta_1), k = 0, 1, \dots. \quad (11)$$

where  $p = p_s$  if  $I = 0$ , and  $p = p_l$  if  $I = 1$ .

Without sacrificing generality, we set  $V_0(\Delta) = \frac{L(\mathbf{a})}{\epsilon p} \Delta$  for all  $\Delta \in \mathcal{S}$ , ensuring that (11) is satisfied at  $k = 0$ . Then, we assume that (11) holds up till  $k > 0$  and hence we have  $V_k(\Delta_2) - V_k(\Delta_1) \geq \frac{L(\mathbf{a})}{\epsilon p}(\Delta_2 - \Delta_1)$  and  $V_k(\Delta_2 + 1) - V_k(\Delta_1 + 1) \geq \frac{L(\mathbf{a})}{\epsilon p}(\Delta_2 - \Delta_1)$ .

Then, we inspect whether it holds for  $k + 1$ . Since  $V_{k+1}(\Delta) = \min_{\mathbf{a} \in \mathcal{A}} Q_k(\Delta, \mathbf{a})$ , we investigate the two state-action value functions, in the following, respectively.

When  $\mathbf{a} = (0, 0)$ , we have

$$\begin{aligned} & \delta Q_k(\Delta_2, \Delta_1) \\ &= Q_k(\Delta_2, (0, 0)) - Q_k(\Delta_1, (0, 0)) \\ &= \Delta_2 - \Delta_1 + \left(1 - \frac{\epsilon}{T_1^u + T_1^c}\right) (V_k(\Delta_2) - V_k(\Delta_1)) + \\ & \quad \frac{\epsilon}{T_1^u + T_1^c} (1 - p_s) (V_k(\Delta_2 + T_1^u + T_1^c) - V_k(\Delta_1 + T_1^u + T_1^c)) \\ &\geq \Delta_2 - \Delta_1 + \left(1 - \frac{\epsilon}{T_1^u + T_1^c}\right) \frac{L(\mathbf{a})}{\epsilon p_s} (\Delta_2 - \Delta_1) \\ &= \frac{L(\mathbf{a})}{\epsilon p} (\Delta_2 - \Delta_1). \end{aligned} \quad (12)$$

When  $\mathbf{a} = (1, 0)$ , we have

$$\begin{aligned} & \delta Q_k(\Delta_2, \Delta_1) \\ &= Q_k(\Delta_2, (1, 0)) - Q_k(\Delta_1, (1, 0)) \\ &= \Delta_2 - \Delta_1 + \left(1 - \frac{\epsilon}{T_2^u + T_1^c}\right) (V_k(\Delta_2) - V_k(\Delta_1)) + \\ & \quad \frac{\epsilon}{T_2^u + T_1^c} (1 - q_s) (V_k(\Delta_2 + T_2^u + T_1^c) - V_k(\Delta_1 + T_2^u + T_1^c)) \\ &\geq \Delta_2 - \Delta_1 + \left(1 - \frac{\epsilon}{T_2^u + T_1^c}\right) \frac{L(\mathbf{a})}{\epsilon q_s} (\Delta_2 - \Delta_1) \end{aligned}$$

$$= \frac{L(\mathbf{a})}{\epsilon p}(\Delta_2 - \Delta_1). \quad (13)$$

When  $\mathbf{a} = (0, 1)$ , we have

$$\begin{aligned} & \delta Q_k(\Delta_2, \Delta_1) \\ &= Q_k(\Delta_2, (0, 0)) - Q_k(\Delta_1, (0, 0)) \\ &= \Delta_2 - \Delta_1 + \left(1 - \frac{\epsilon}{T_1^u + T_2^c}\right) (V_k(\Delta_2) - V_k(\Delta_1)) + \\ & \quad \frac{\epsilon}{T_1^u + T_2^c} (1 - p_l) (V_k(\Delta_2 + T_1^u + T_2^c) - V_k(\Delta_1 + T_1^u + T_2^c)) \\ &\geq \Delta_2 - \Delta_1 + \left(1 - \frac{\epsilon}{T_1^u + T_2^c}\right) \frac{L(\mathbf{a})}{\epsilon p_l} (\Delta_2 - \Delta_1) \\ &= \frac{L(\mathbf{a})}{\epsilon p} (\Delta_2 - \Delta_1). \end{aligned} \quad (14)$$

When  $\mathbf{a} = (1, 1)$ , we have

$$\begin{aligned} & \delta Q_k(\Delta_2, \Delta_1) \\ &= Q_k(\Delta_2, (0, 0)) - Q_k(\Delta_1, (0, 0)) \\ &= \Delta_2 - \Delta_1 + \left(1 - \frac{\epsilon}{T_2^u + T_2^c}\right) (V_k(\Delta_2) - V_k(\Delta_1)) + \\ & \quad \frac{\epsilon}{T_2^u + T_2^c} (1 - q_l) (V_k(\Delta_2 + T_2^u + T_2^c) - V_k(\Delta_1 + T_2^u + T_2^c)) \\ &\geq \Delta_2 - \Delta_1 + \left(1 - \frac{\epsilon}{T_2^u + T_2^c}\right) \frac{L(\mathbf{a})}{\epsilon q_l} (\Delta_2 - \Delta_1) \\ &= \frac{L(\mathbf{a})}{\epsilon p} (\Delta_2 - \Delta_1). \end{aligned} \quad (15)$$

This concludes the proof of Lemma 3.

#### D. Proof of Theorem 1

First, we introduce  $\delta Q(\Delta_2, \Delta_1, \mathbf{a}) = Q(\Delta_2, \mathbf{a}) - Q(\Delta_1, \mathbf{a})$  for convenience. For any  $\Delta_1, \Delta_2 \in \mathcal{S}$ , and  $\Delta_1 \leq \Delta_2$ , we have

$$\begin{aligned} & \delta Q(\Delta_2, \Delta_1, \hat{\mathbf{a}}) - (V(\Delta_2) - V(\Delta_1)) \\ &= \Delta_2 - \Delta_1 - \frac{\epsilon}{L(\hat{\mathbf{a}})} (V(\Delta_2) - V(\Delta_1)) \\ & \quad + \frac{\epsilon(1-p)}{L(\hat{\mathbf{a}})} (V(\Delta_2 + L(\hat{\mathbf{a}})) - V(\Delta_1 + L(\hat{\mathbf{a}}))). \end{aligned} \quad (16)$$

Given that the concavity of  $V(s)$  is established in Lemma 2, it follows that  $V(\Delta_2 + L(\hat{\mathbf{a}})) - V(\Delta_1 + L(\hat{\mathbf{a}})) \leq V(\Delta_2) - V(\Delta_1)$ .

Then, we can get that

$$\begin{aligned} & \delta Q(\Delta_2, \Delta_1, \hat{\mathbf{a}}) - (V(\Delta_2) - V(\Delta_1)) \\ &\leq \Delta_2 - \Delta_1 + \frac{\epsilon(1-p)}{L(\hat{\mathbf{a}})} (V(\Delta_2) - V(\Delta_1)) \\ & \quad - \frac{\epsilon}{L(\hat{\mathbf{a}})} (V(\Delta_2) - V(\Delta_1)) \end{aligned}$$

$$= \Delta_2 - \Delta_1 - \frac{\epsilon p}{L(\hat{\mathbf{a}})}(V(\Delta_2) - V(\Delta_1)). \quad (17)$$

As shown in Lemma 3, we have  $V(\Delta_2) - V(\Delta_1) \geq \frac{L(\hat{\mathbf{a}})}{\epsilon p}(\Delta_2 - \Delta_1)$ . This implies that  $\Delta Q(\Delta_2, \Delta_1) - (V(\Delta_2) - V(\Delta_1)) \leq 0$ .

Next, we prove the threshold structure of the optimal policy. Suppose  $\Delta_2 \geq \Delta_1$  and  $\pi^*(\Delta_1) = \hat{\mathbf{a}}$ , we have  $V(\Delta_1) = Q(\Delta_1, \hat{\mathbf{a}})$ , i.e.,  $V(\Delta_1) = Q(\Delta_1, \hat{\mathbf{a}})$ . It is straightforward to obtain  $V(\Delta_2) \geq Q(\Delta_2, \hat{\mathbf{a}})$ , since  $V(\Delta_2) - V(\Delta_1) \geq Q(\Delta_2, \hat{\mathbf{a}}) - Q(\Delta_1, \hat{\mathbf{a}})$ . Moreover, since the value function is a minimum of two state-action value functions, we have  $V(\Delta_2) \leq Q(\Delta_2, \hat{\mathbf{a}})$ . Therefore, we can conclude that  $V(\Delta_2) = Q(\Delta_2, \hat{\mathbf{a}})$  and that  $\pi^*(\Delta_2) = \hat{\mathbf{a}}$ .

This completes the proof of Theorem 1.

## REFERENCES

- [1] P. Bertsekas, Dimitri, *Dynamic Programming and Optimal Control-II*, 3rd ed. Belmont, MA, USA: Athena Sci., 2007, vol. 2.