Definition 1. The two reward functions $\mathcal{R}_1: S^{\otimes} \times 2^{E^{\otimes}} \to \mathbb{R}$ and $\mathcal{R}_2: S^{\otimes} \times E^{\otimes} \times S^{\otimes} \to \mathbb{R}$ are defined as follows.

$$\mathcal{R}_1(s^{\otimes}, \pi) = \begin{cases} r_n |\pi| & \text{if } [s^{\otimes}]_q \notin SinkSet, \\ 0 & \text{otherwise,} \end{cases}$$
 (1)

where |E| means number of elements in the set E and r_n is a positive value.

$$\mathcal{R}_{2}(s^{\otimes}, e, s^{\otimes'}) = \begin{cases}
r_{p} & \text{if } \exists i \in \{1, \dots, n\}, \ (s^{\otimes}, e, s^{\otimes'}) \in \bar{F}_{i}^{\otimes}, \\
r_{sink} & \text{if } [s^{\otimes'}]_{q} \in SinkSet, \\
0 & \text{otherwise,}
\end{cases} \tag{2}$$

where r_p and r_{sink} are the positive and negative value, respectively.

For a Markov chain MC_{SV}^{\otimes} induced by a product MDP D^{\otimes} with a supervisor SV, let $S_{SV}^{\otimes} = T_{SV}^{\otimes} \sqcup R_{SV}^{\otimes 1} \sqcup \ldots \sqcup R_{SV}^{\otimes h}$ be the set of states in MC_{SV}^{\otimes} , where T_{SV}^{\otimes} is the set of transient states and $R_{SV}^{\otimes i}$ is the recurrent class for each $i \in \{1, \ldots, h\}$, and let $R(MC_{SV}^{\otimes})$ be the union of all recurrent classes in MC_{SV}^{\otimes} . Let $\delta_{SV,i}^{\otimes}$ be the set of transitions in a recurrent class $R_{SV}^{\otimes i}$, namely $\delta_{SV,i}^{\otimes} = \{(s^{\otimes}, e, s^{\otimes'}) \in \delta^{\otimes}; s^{\otimes} \in R_{SV}^{\otimes i}, \ P_T^{\otimes}(s^{\otimes'}|s^{\otimes}, e) > 0, P_E^{\otimes}(e|s^{\otimes}, SV(s^{\otimes})) > 0\}$, and let $P_{SV}^{\otimes} : S_{SV}^{\otimes} \times S_{SV}^{\otimes} \to [0, 1]$ such that $P_{SV}^{\otimes}(s^{\otimes'}|s^{\otimes}) = \sum_{e \in SV(s^{\otimes})} P_T^{\otimes}(s^{\otimes'}|s^{\otimes}, e) P_E^{\otimes}(e|s^{\otimes}, SV(s^{\otimes}))$ be the transition probability under SV.

Lemma 1. For any supervisor SV and any recurrent class $R_{SV}^{\otimes i}$ in the Markov chain MC_{SV}^{\otimes} , MC_{SV}^{\otimes} satisfies one of the following conditions.

1.
$$\delta_{SV,i}^{\otimes} \cap \bar{F}_j^{\otimes} \neq \emptyset$$
, $\forall j \in \{1, \dots, n\}$,

2.
$$\delta_{SV,i}^{\otimes} \cap \bar{F}_j^{\otimes} = \emptyset$$
, $\forall j \in \{1, \dots, n\}$.

Definition 2. An accepting recurrent class is defined as the recurrent class whose at least one accepting transition in each accepting set \bar{F}_j^{\otimes} with $j \in \{1, \ldots, n\}$.

Theorem 1. Let M^{\otimes} be the product DES corresponding to a DES M and an LTL formula φ . Let \mathcal{R}_1 be a reward function for control patterns. If there exists a supervisor SV satisfying φ and it satisfies that there is no state $s^{\otimes} \in S_{SV}^{\otimes}$ reachable from initial state s_{init}^{\otimes} such that $[\![s^{\otimes}]\!]_q \in SinkSet$, then there exist a discount factor γ^* , a positive reward $r_p^*(\mathcal{R}_1)$ that is a function of \mathcal{R}_1 and satisfies $r_p^*(\mathcal{R}_1) >> ||\mathcal{R}_1||_{\infty}$, and a negative reward $r_{sink}^*(r_p, \mathcal{R}_1)$ that is a function of r_p and \mathcal{R}_1 and satisfies $r_{sink}(r_p, \mathcal{R}_1) << -(r_p+||\mathcal{R}_1||_{\infty})$ such that any algorithm that maximizes the expected discounted reward with $\gamma > \gamma^*$, $r_p > r_p^*(\mathcal{R}_1)$, and $r_{sink} < r_{sink}^*(r_p^*, \mathcal{R}_1)$ will find, with probability one, a supervisor satisfying φ and it satisfies that there is no state $s^{\otimes} \in S_{SV}^{\otimes}$ reachable from the initial state s_{init}^{\otimes} such that $[\![s^{\otimes}]\!]_q \in SinkSet$.

Proof. Suppose that SV^* be an optimal supervisor but does not satisfy the LTL formula φ or there is a state s_{sink}^{\otimes} reachable from the initial state such that $[s_{sink}^{\otimes}]_q \in SinkSet$ under the supervisor SV^* . Then, for any recurrent class $R_{SV^*}^{\otimes i}$ in the Markov chain $MC_{SV^*}^{\otimes}$ and any accepting set \bar{F}_j^{\otimes} of the product DES M^{\otimes} , $\delta_{SV^*,i}^{\otimes} \cap \bar{F}_j^{\otimes} = \emptyset$ holds for the first case by Lemma 1 and there is a recurrent class $R_{SV^*}^{\otimes i}$ such that $s_{sink}^{\otimes} \in R_{SV^*}^{\otimes i}$ for the second case. We consider the two cases separately.

1. Assume that SV^* does not the LTL formula φ . By the assumption, the system under the supervisor SV^* can obtain rewards only in the set of transient states. We consider the best scenario in the assumption. Let $p^k(s,s')$ be the probability of going to a state s' in k time steps after leaving the state s, and let $Post(T_{SV^*}^{\otimes})$ be the set of states in recurrent classes that can be transitioned from states in $T_{SV^*}^{\otimes}$ by one event occurrence. For the initial state s_{init}^{\otimes} in the set of transient states, it holds that

$$V^{SV^*}(s_{init}^{\otimes}) = \sum_{k=0}^{\infty} \sum_{s^{\otimes} \in T_{SV^*}^{\otimes}} \gamma^k p^k(s_{init}^{\otimes}, s^{\otimes})$$

$$\sum_{s^{\otimes'} \in T_{SV^*}^{\otimes} \cup Post(T_{\pi^*}^{\otimes})} \sum_{e \in SV(s^{\otimes})} P_T^{\otimes}(s^{\otimes'}|s^{\otimes}, e) P_E^{\otimes}(e|s^{\otimes}, SV(s^{\otimes})) \mathcal{R}(s^{\otimes}, SV(s^{\otimes}), e, s^{\otimes'})$$

$$\leq r_p \sum_{k=0}^{\infty} \sum_{s^{\otimes} \in T_{SV^*}^{\otimes}} \gamma^k p^k(s_{init}^{\otimes}, s^{\otimes}) + \sum_{k=0}^{\infty} \gamma^k ||\mathcal{R}_1||_{\infty}.$$

By the property of the transient states, for any state s^{\otimes} in $T_{SV^*}^{\otimes}$, there exists a bounded positive value m such that $\sum_{k=0}^{\infty} \gamma^k p^k(s_{init}^{\otimes}, s^{\otimes}) \leq \sum_{k=0}^{\infty} p^k(s_{init}^{\otimes}, s^{\otimes}) < m$ [1]. Therefore, there exists a bounded positive value \bar{m} such that $V^{SV^*}(s_{init}^{\otimes}) < \bar{m} + \frac{1}{1-\gamma}||\mathcal{R}_1||_{\infty}$.

2. Assume that there is a state s_{sink}^{\otimes} reachable from the initial state such that $\llbracket s_{sink}^{\otimes}
Vert_q \in SinkSet$ under SV^* . By the assumption, there is a recurrent class $R_{SV^*}^{\otimes i}$ reachable from the initial state such that $s_{sink}^{\otimes} \in R_{SV^*}^{\otimes i}$. We consider the best scenario in the assumption. We assume that all of the recurrent classes except for $R_{SV^*}^{\otimes i}$ are the accepting recurrent classes and there exist a number l>0, a state s_{sink}^{\otimes} in $Post(T_{SV^*}^{\otimes}) \cap R_{SV^*}^{\otimes i}$, and a subset of transient states $\{s_1^{\otimes},\ldots,s_{l-1}^{\otimes}\} \subset T_{SV^*}^{\otimes}$ such that $p(s_{init}^{\otimes},s_1^{\otimes})>0$, $p(s_i^{\otimes},s_{i+1}^{\otimes})>0$ for $i\in\{1,\ldots,l-2\}$, and $p(s_{l-1}^{\otimes},s_{sink}^{\otimes})>0$ by the property of transient states. We have

$$V^{SV^*}(s_{init}^{\otimes}) < Pr_{SV^*}^{M^{\otimes}}(s_{init}^{\otimes} \models \varphi) \sum_{k=0}^{\infty} \gamma^k (r_p + ||\mathcal{R}_1||_{\infty}) + \gamma^l p^l(s_{init}^{\otimes}, s_{sink}^{\otimes}) \sum_{k=0}^{\infty} \gamma^k r_{sink}$$

$$+Pr_{SV^*}^{M^{\otimes}}(s_{init}^{\otimes} \not\models \varphi)(r_p + ||\mathcal{R}_1||_{\infty}) \sum_{k=0}^{\infty} \sum_{s^{\otimes} \in T_{\pi^*}^{\otimes}} \gamma^k p^k(s_{init}^{\otimes}, s^{\otimes})$$

$$< \frac{1}{1-\gamma} \{ Pr_{SV^*}^{M^{\otimes}}(s_{init}^{\otimes} \models \varphi)(r_p + ||\mathcal{R}_1||_{\infty}) + \gamma^l p^l(s_{init}^{\otimes}, s_{sink}^{\otimes}) r_{sink} \} + \bar{m}',$$

where \bar{m}' is a constant such that $\bar{m}' > Pr_{SV^*}^{M^{\otimes}}(s_{init}^{\otimes} \not\models \varphi)(r_p + ||\mathcal{R}_1||_{\infty}) \sum_{k=0}^{\infty} \sum_{s^{\otimes} \in T_{\pi^*}^{\otimes}} \gamma^k p^k(s_{init}^{\otimes}, s^{\otimes}).$ Therefore, if it holds that $r_{sink} \leq -\frac{Pr_{SV^*}^{M^{\otimes}}(s_{init}^{\otimes} \models \varphi)}{\gamma^l p^l(s_{init}^{\otimes}, s_{sink}^{\otimes})}(r_p + ||\mathcal{R}_1||_{\infty}),$ we then have $V^{SV^*}(s_{init}^{\otimes}) < \bar{m}'$ for any $\gamma \in [0, 1)$.

Let SV be a supervisor satisfying φ and it satisfies that there is no state $s^{\otimes} \in S_{SV}^{\otimes}$ reachable from initial state s_{init}^{\otimes} such that $[\![s^{\otimes}]\!]_q \in SinkSet$. We consider the following two cases.

1. Assume that the initial state s_{init}^{\otimes} is in a recurrent class $R_{S\bar{V}}^{\otimes i}$ for some $i \in \{1, \ldots, h\}$. For any accepting set \bar{F}_j^{\otimes} , $\delta_{S\bar{V}, i}^{\otimes} \cap \bar{F}_j^{\otimes} \neq \emptyset$ holds by the definition of $S\bar{V}$. The expected discounted reward for s_{init}^{\otimes} is given by

$$V^{\bar{SV}}(s_{init}^{\otimes}) = \mathbb{E}^{SV}[\sum_{k=0}^{\infty} \gamma^k \mathcal{R}(s_k, \pi_k, e_k, s_{k+1}) | s_0 = s_{init}^{\otimes}]$$
 (3)

Since s_{init}^{\otimes} is in $R_{\pi}^{\otimes i}$, there exists a set of positive numbers $K = \{k \; ; \; k \geq n, p^k(s_{init}^{\otimes}, s_{init}^{\otimes}) > 0\}$ [1]. We consider the worst scenario of returning the initial state in this case. For the stopping time k of first returning to the initial state, it holds that

$$\begin{split} V^{\bar{SV}}(s_{init}^{\otimes}) > & \mathbb{E}^{\bar{SV}}[\gamma^{k-1}r_p + \gamma^{k-1}V^{\bar{SV}}(s_{init}^{\otimes})|s_0 = s_{init}^{\otimes}] \\ \geq & \gamma^{\mathbb{E}^{\bar{SV}}[k-1|s_0 = s_{init}^{\otimes}]}r_p + \gamma^{\mathbb{E}^{\bar{SV}}[k-1|s_0 = s_{init}^{\otimes}]}V^{\bar{SV}}(s_{init}^{\otimes}) \\ = & \frac{\gamma^{\mathbb{E}^{\bar{SV}}[k-1|s_0 = s_{init}^{\otimes}]}r_p}{1 - \gamma^{\mathbb{E}^{\bar{SV}}[k-1|s_0 = s_{init}^{\otimes}]}}, \end{split}$$

where second inequality holds since it holds that $\mathbb{E}^{\bar{SV}}[\gamma^k|s_0=s_{init}^{\otimes}] \geq \gamma^{\mathbb{E}^{SV}[k|s_0=s_{init}^{\otimes}]}$ by Jensen's inequality. If it holds that $\gamma^{\mathbb{E}^{SV}[k-1|s_0=s_{init}^{\otimes}]}r_p^* > (1+\ldots+\gamma^{\mathbb{E}^{SV}[k|s_0=s_{init}^{\otimes}]})||\mathcal{R}_1||_{\infty}$, we then have $\frac{\gamma^{\mathbb{E}^{SV}[k-1|s_0=s_{init}^{\otimes}]}}{1-\gamma^{\mathbb{E}^{SV}[k-1|s_0=s_{init}^{\otimes}]}}r_p > \frac{1}{1-\gamma}||\mathcal{R}_1||_{\infty}$. Therefore, for a reward function \mathcal{R}_1 , there exist $\gamma^* < 1$, a positive reward r_p^* that satisfies $\gamma^{\mathbb{E}^{\bar{SV}}[k-1|s_0=s_{init}^{\otimes}]}r_p^* > (1+\ldots+\gamma^{\mathbb{E}^{\bar{SV}}[k|s_0=s_{init}^{\otimes}]})||\mathcal{R}_1||_{\infty}$, and a negative reward r_{sink} that satisfies $r_{sink} \leq -\frac{Pr_{SV^*}^{M^{\otimes}}(s_{init}^{\otimes})+\varphi}{\gamma^l p^l(s_{init}^{\otimes},s_{sink}^{\otimes})}}(r_p+||\mathcal{R}_1||_{\infty})$ such that $\gamma>\gamma^*$, $r_p>r_p^*$, and $r_{sink}^* < r_{sink}$ imply $V^{\bar{SV}}(s_{init}^{\otimes}) > V^{SV^*}(s_{init}^{\otimes})$.

2. Assume that the initial state s_{init}^{\otimes} is in the set of transient states $T_{SV}^{\otimes}.P_{SV}^{M^{\otimes}}(s_{init}^{\otimes} \models \varphi) > 0$ holds by the definition of SV. For a recurrent class $R_{SV}^{\otimes i}$ such that $\delta_{SV,i}^{\otimes} \cap \bar{F}_{j}^{\otimes} \neq \emptyset$ for each accepting set \bar{F}_{j}^{\otimes} , there exist a number l' > 0, a state \hat{s}^{\otimes} in $Post(T_{SV}^{\otimes}) \cap R_{SV}^{\otimes i}$, and a subset of transient states $\{s_{1}^{\otimes},\ldots,s_{l'-1}^{\otimes}\} \subset T_{SV}^{\otimes}$ such that $p(s_{init}^{\otimes},s_{1}^{\otimes}) > 0$, $p(s_{i}^{\otimes},s_{i+1}^{\otimes}) > 0$ for $i \in \{1,\ldots,l'-2\}$, and $p(s_{l'-1}^{\otimes},\hat{s}^{\otimes}) > 0$ by the property of transient states. Hence, it holds that $p'(s_{init}^{\otimes},\hat{s}^{\otimes}) > 0$ for the state \hat{s}^{\otimes} . Thus, for the stopping time k of first returning to the state \hat{s}^{\otimes} , by ignoring positive rewards in T_{SV}^{\otimes} , we have

$$\begin{split} &V^{S\overline{V}}(s_{init}^{\otimes}) \\ =& \mathbb{E}^{SV}[\sum_{m=0}^{\infty} \gamma^m \mathcal{R}(s_m, S\overline{V}(s_m), e_m, s_{m+1}) | s_0 = s_{init}^{\otimes}] \\ \geq& \mathbb{E}^{SV}[\gamma^l \sum_{m=0}^{\infty} \gamma^m \mathcal{R}(s_{m+l}, S\overline{V}(s_{m+l}), e_{m+l}, s_{m+l+1}) | s_0 = s_{init}^{\otimes}] \\ \geq& \gamma^{l'} p^{l'}(s_{init}^{\otimes}, \hat{s}^{\otimes}) \mathbb{E}^{S\overline{V}}[\gamma^{k-1} r_p + \gamma^{k-1} V^{S\overline{V}}(\hat{s}^{\otimes}) | s_{l'} = \hat{s}^{\otimes}] \\ \geq& \gamma^{l'} p^{l'}(s_{init}^{\otimes}, \hat{s}^{\otimes}) \{\gamma^{\mathbb{E}^{S\overline{V}}[k-1|s_{l'}=\hat{s}^{\otimes}]} r_p + \gamma^{\mathbb{E}^{S\overline{V}}[k-1|s_{l'}=\hat{s}^{\otimes}]} V^{S\overline{V}}(\hat{s}^{\otimes})\} \\ =& \gamma^{l'} p^{l'}(s_{init}^{\otimes}, \hat{s}^{\otimes}) \frac{\gamma^{\mathbb{E}^{S\overline{V}}[k-1|s_{l'}=\hat{s}^{\otimes}]} r_p}{1 - \gamma^{\mathbb{E}^{S\overline{V}}[k-1|s_{l'}=\hat{s}^{\otimes}]}}. \end{split}$$

If it holds that $\gamma^l p^l(s_{init}^{\otimes}, \hat{s}^{\otimes}) \gamma^{\mathbb{E}^{SV}[k-1|s_0=s_{init}^{\otimes}]} r_p > (1+\ldots+\gamma^{\mathbb{E}^{SV}[k|s_0=s_{init}^{\otimes}]}) ||\mathcal{R}_1||_{\infty}$, we then have $\gamma^{l'} p^{l'}(s_{init}^{\otimes}, \hat{s}^{\otimes}) \frac{\gamma^{\mathbb{E}^{SV}[k-1|s_{l'}=\hat{s}^{\otimes}]}}{1-\gamma^{\mathbb{E}^{SV}[k-1|s_{l'}=\hat{s}^{\otimes}]}} r_p > \frac{1}{1-\gamma} ||\mathcal{R}_1||_{\infty}$ for any $\gamma \in [0,1)$. Therefore, for the reward function \mathcal{R}_1 , there exist $\gamma^* < 1$, a positive reward r_p^* that satisfies $\gamma^l p^l(s_{init}^{\otimes}, \hat{s}^{\otimes}) \gamma^{\mathbb{E}^{SV}[k-1|s_l=\hat{s}^{\otimes}]} r_p^* > (1+\ldots+\gamma^{\mathbb{E}^{SV}[k|s_l=\hat{s}^{\otimes}]}) ||\mathcal{R}_1||_{\infty}$, and a negative reward r_{sink} that satisfies $r_{sink} \leq -\frac{Pr_{SV^*}^{M^{\otimes}}(s_{init}^{\otimes}|=\varphi)}{\gamma^l p^l(s_{init}^{\otimes}, s_{sink}^{\otimes})} (r_p + ||\mathcal{R}_1||_{\infty})$ such that $\gamma > \gamma^*$, $r_p > r_p^*$, and $r_{sink}^* < r_{sink}$ imply $V^{SV}(s_{init}^{\otimes}) > V^{SV^*}(s_{init}^{\otimes})$.

The results contradict the optimality assumption of SV^*

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