

## Problem 1

(a)

$$\begin{cases} -2x_1 + x_2 + x_3^3 = 0 \\ x_1 - x_2 - x_3^3 = 0 \\ -x_3 - x_3^3 = 0 \end{cases}$$

$$x_3^3 = 2x_1 - x_2 = x_1 - x_2 = -x_3$$

$$x_1 = x_2 = x_3 = 0$$

(b)

$$\begin{aligned} V(\mathbf{x}) &= x_1^2 + 2x_2^2 + 2x_3^2 + 2x_1x_2 - 2x_2x_3 \\ &= (x_1^2 + 2x_1x_2 + x_2^2) + (x_2^2 - 2x_2x_3 + x_3^2) + x_2^2 + x_3^2 \\ &= (x_1 + x_2)^2 + (x_2 - x_3)^2 + x_2^2 + x_3^2 \end{aligned}$$

Thus we know  $V(\mathbf{x}) = 0$  when  $x_1 = x_2 = x_3 = 0$ , and  $V(\mathbf{x}) \geq 0$ . To prove that  $V(\mathbf{x}) > 0$  when  $\mathbf{x} \neq \mathbf{0}$ , consider a hypothetical contradicting case where  $\mathbf{x} \neq \mathbf{0}$  and  $V(\mathbf{x}) = 0$ ; in this case since all terms in the sum are non-negative, all terms must be exactly 0; therefore we know immediately that  $x_2 = x_3 = 0$ . Since  $x_1 + x_2 = 0$  and  $x_2 = 0$ ,  $x_1 = 0$ , so we have  $x_1 = x_2 = x_3 = 0$ . So it is not possible to construct the contradicting case, and thus  $V(\mathbf{x})$  must be strictly greater than 0 when  $\mathbf{x} \neq \mathbf{0}$ .

(c)

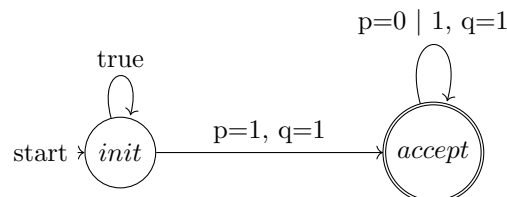
$$\frac{\partial V}{\partial \mathbf{x}} = \begin{bmatrix} 2x_1 + 2x_2 \\ 4x_2 + 2x_1 - 2x_3 \\ 4x_3 - 2x_2 \end{bmatrix}$$

$$\begin{aligned} \frac{\partial V}{\partial \mathbf{x}} \cdot \begin{bmatrix} -2x_1 + x_2 + x_3^3 \\ x_1 - x_2 - x_3^3 \\ -x_3 - x_3^3 \end{bmatrix} &= (2x_1 + 2x_2) * (-2x_1 + x_2 + x_3^3) + (4x_2 + 2x_1 - 2x_3) * (x_1 - x_2 - x_3^3) \\ &\quad + (4x_3 - 2x_2) * (-x_3 - x_3^3) \\ &= -2x_3^4 - 4x_3^2 - 2x_1x_3 + 4x_2x_3 - 2x_1^2 - 2x_2^2 \\ &= -(x_1^2 + 2x_1x_3 + x_3^2) - 2(x_2^2 - 2x_2x_3 + x_3^2) - x_1^2 - x_3^2 - 2x_3^4 \\ &= -(x_1 + x_3)^2 - 2(x_2 - x_3)^2 - x_1^2 - x_3^2 - 2x_3^4 \end{aligned}$$

All square and 4th-order terms must be non-negative, so the sum of their negations must be non-positive. And it's obvious that the equation equals 0 when  $x_1 = x_2 = x_3 = 0$ .

## Problem 2

(a)



(If  $q = 0$  while at the accepting state, there will be no corresponding enabled transition so the automaton will no longer accept.)

- (b) 1. False;  $(pq)^+ \models \mathbf{G}(p \vee q)$  but  $(pq)^+ \not\models \mathbf{G}(p) \vee \mathbf{G}(q)$ .<sup>1</sup>
2. True;  $\mathbf{G}(p) \vee \mathbf{G}(q) \equiv \forall m \geq 0, (\exists a \geq m : \pi, a \models p) \vee (\exists b \geq m : \pi, b \models q) \implies \forall m \geq 0, \exists n \geq m : \pi, n \models (\text{set } n \text{ to either } a \text{ or } b) (p \vee q) \equiv \mathbf{GF}(p \vee q)$ ; the reverse also holds since  $\mathbf{GF}(p \vee q) \equiv \forall m \geq 0, \exists n \geq m : \pi, n \models (p \vee q) \implies \forall m \geq 0, (\exists a \geq m : \pi, a \models p) \vee (\exists b \geq m : \pi, b \models q)$  (set both  $a$  and  $b$  to  $n$ )  $\equiv \mathbf{G}(p) \vee \mathbf{G}(q)$ .
3. True;  $(p \wedge q)\mathbf{U}r \equiv (\exists m \geq 0 : \pi, m \models r) \wedge (\forall n \text{ s.t. } 0 \leq n < m, \pi, n \models (p \wedge q)) \iff (\exists m \geq 0 : \pi, m \models r) \wedge (\forall a \text{ s.t. } 0 \leq a < m, \pi, a \models p) \wedge (\forall b \text{ s.t. } 0 \leq b < m, \pi, b \models q)$  ( $a$  and  $b$  span the same  $\forall$  range as  $n$ , so the two can be combined to give  $p \wedge q$  and vice versa)  $\equiv (p\mathbf{U}r) \wedge (q\mathbf{U}r)$ .
- (c) Xiao's right, i.e. the automaton accepts traces that the formula cannot accept.  $\mathbf{F}(p \wedge q)$  enforces that  $\exists m \geq 0 : (p = 1 \wedge q = 1)$ , but this is not enforced by the automaton; for example  $qqqp^+$  can be accepted by the automaton, but cannot be accepted by the formula.<sup>2</sup>

### Problem 3

(a)

	0	0.5	1.0	1.5	2.0	2.5	3.0	3.5
$\mu = x - 0$	1	0	1	1	1	0	0	0
$\mathbf{F}_{[0,0.5]}\mu$	1	1	1	1	1	0	0	
$\mathbf{G}_{[[0,0.5]]}\mathbf{F}_{[0,0.5]}\mu$	1	1	1	1	0	0		

$\rho(\varphi, \mathbf{x}, 0) = 1 > 0$ , and moreover,  $\rho(\varphi, \mathbf{x}, [0 : 2.5]) \geq 0^3$ , so  $x \models \varphi$  for the interval  $[0 : 2.5]$ .

(b)

	0	0.5	1.0	1.5	2.0	2.5	3.0	3.5
$\mu_1 = 0 - y$	-0.1	-0.3	0.1	3	0.1	-1	-2	0.1
$\mathbf{F}_{[1,1.5]}\mu_1$	3	3	0.1	-1	0.1			
$\mu_2 = x - 0.5$	0.5	-0.5	0.5	0.5	0.5	-0.5	-0.5	-0.5
$\neg\mu_2$	-0.5	0.5	-0.5	-0.5	-0.5	0.5	0.5	0.5
$\neg\mu_2 \vee \mathbf{F}_{[1,1.5]}\mu_1$	3	3	0.1	-0.5	0.1			
$\mathbf{G}_{[0,2]}(\neg\mu_2 \vee \mathbf{F}_{[1,1.5]}\mu_1)$	-0.5							

$\rho(\psi, \mathbf{x}, 0) = -0.5 < 0$ , so  $\mathbf{x} \not\models \psi$ .

(c) Make  $x = 0.5$  at  $t = 1.5$ .

(d) Yes; the violations are caused by extreme values that are on the two ends of the interpolation; values in the linear interpolation cannot change the extreme values, so violations cannot be undone.

<sup>1</sup> $(pq)^+$  means  $p = 1, q = 0$ , then  $p = 0, q = 1$ , and repeat this pattern infinitely.

<sup>2</sup> $qqqp^+$  means  $q=1$  and  $p=0$  for the first 3 steps, then  $q=0$  and  $p=1$  for all following steps.

<sup>3</sup> $\rho = 0$  is fine here since the requirement is  $x \geq 0$ .

## Problem 4

- (a) We are looking for probabilities of being in either  $q_0$  or  $q_3$  at time 0, 1 and 2.

At time 0, since initial probability distribution is assumed to be uniform, there's a 0.25 probability for being in each of  $q_0, q_1, q_2, q_3$ .  $0.25 + 0.25 = 0.5$ .

At time 1,  $P(q_0) = \sum_q P(s_1 = q_0, s_0 = q) = \sum_q P(s_0 = q) * p(q, q_0) = \sum_q 0.25 * p(q, q_0)$ , and same for  $q_1, q_2, q_3$ , thus:

$$P(q_0) = 0.25 * 0.1 + 0.25 * 0.5 + 0.25 * 0 + 0.25 * 0.3 = 0.225$$

$$P(q_1) = 0.25 * 0.2 + 0.25 * 0.5 + 0.25 * 0.9 + 0.25 * 0.2 = 0.45$$

$$P(q_2) = 0.25 * 0.3 + 0.25 * 0 + 0.25 * 0.1 + 0.25 * 0.5 = 0.225$$

$$P(q_3) = 0.25 * 0.4 + 0.25 * 0 + 0.25 * 0 + 0.25 * 0 = 0.1$$

$$0.225 + 0.1 = 0.325.$$

At time 2, we do the same calculations and get:

$$P(q_0) = 0.225 * 0.1 + 0.45 * 0.5 + 0.225 * 0 + 0.1 * 0.3 = 0.2775$$

$$P(q_1) = 0.225 * 0.2 + 0.45 * 0.5 + 0.225 * 0.9 + 0.1 * 0.2 = 0.4925$$

$$P(q_2) = 0.225 * 0.3 + 0.45 * 0 + 0.225 * 0.1 + 0.1 * 0.5 = 0.14$$

$$P(q_3) = 0.225 * 0.4 + 0.45 * 0 + 0.225 * 0 + 0.1 * 0 = 0.09$$

$$0.2775 + 0.09 = 0.3675.$$

- (b) We are looking for the probability of being in either  $q_1$  or  $q_2$  at time 2 and not in state  $q_1$  or  $q_3$  at time 0 and 1. Later time steps are dependent on previous time steps, so we have to use conditional probabilities.  $P(s_2 = q_1 \text{ or } q_2, s_1 = q_0 \text{ or } q_2, s_0 = q_0 \text{ or } q_2) = P(s_2 = q_1 \text{ or } q_2 \mid s_1 = q_0 \text{ or } q_2, s_0 = q_0 \text{ or } q_2)P(s_1 = q_0 \text{ or } q_2 \mid s_0 = q_0 \text{ or } q_2)P(s_0 = q_0 \text{ or } q_2)$ .

$$P(s_0 = q_0 \text{ or } q_2) = 0.5.$$

$$P(s_1 = q_0 \text{ or } q_2 \mid s_0 = q_0 \text{ or } q_2) = \frac{P(s_1=q_0 \text{ or } q_2, s_0=q_0 \text{ or } q_2)}{P(s_0=q_0 \text{ or } q_2)} = \frac{0.25*0.1+0.25*0.3+0.25*0.0+0.25*0.1}{0.5} = 0.25.$$

$$P(s_2 = q_1 \text{ or } q_2 \mid s_1 = q_0 \text{ or } q_2, s_0 = q_0 \text{ or } q_2) = P(s_2 = q_1 \text{ or } q_2 \mid s_1 = q_0 \text{ or } q_2) = \frac{P(s_2=q_1 \text{ or } q_2, s_1=q_0 \text{ or } q_2)}{P(s_1=q_0 \text{ or } q_2)} = \frac{0.225*0.2+0.225*0.3+0.225*0.9+0.225*0.1}{0.225+0.225} = 0.75.$$

$$0.5 * 0.25 * 0.75 = 0.09375.$$

## Problem 5

- (a)  $(left, x = -1) \xrightarrow{\delta t=7} (left, x = 2.5)$   
 $(left, x = 2.5) \xrightarrow{2.5>2} (right, x = 2.5)$   
 $(right, x = 2.5) \xrightarrow{\delta t=5} (right, x = -2.5)$   
 $(right, x = 2.5) \xrightarrow{-2.5<-2} (right, x = -2.5)$

- (b) The largest possible value of  $x$  when entering *left* is  $-2$ ; the smallest is  $-4$ . The largest possible value of  $x$  when leaving *left* is  $4$ ; the smallest is  $2$ .

$$\text{Maximum dwell time: } \frac{4-(-4)}{0.5} = 16.$$

$$\text{Minimum dwell time: } \frac{2-(-2)}{0.5} = 8.$$

## Problem 6

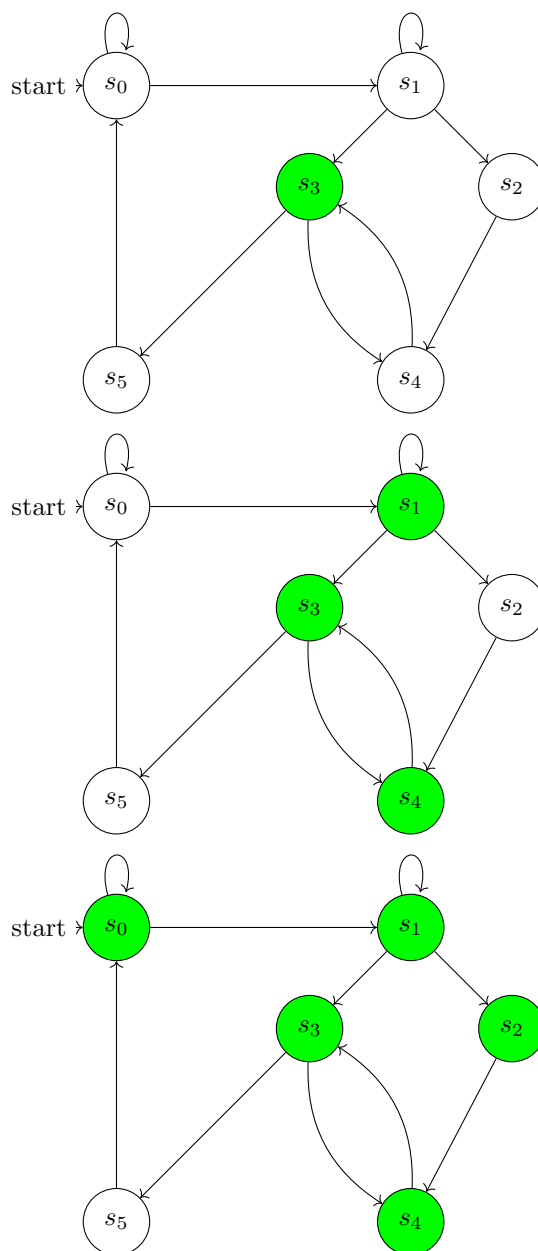
- (a) No. Although the base case is trivially true because of initialization, we can find the counter-example  $i' = 2, j' = 3, x' = true$  for the inductive case, which satisfies the expression since the precondition is

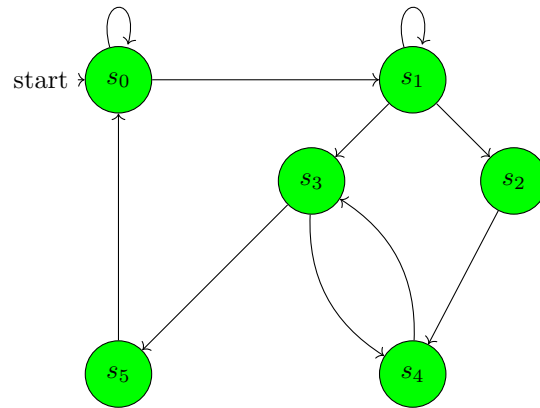
false; in this case only T2 is enabled, so  $(j' - 1) = 2$  while  $i', x'$  don't change and the precondition becomes true; however the postcondition  $x = true \neq false$  is false, so the entire expression becomes false. Since it's not an inductive invariant, it cannot be used to prove safety invariant.

- (b) No. Since  $(j = 0)$  has to be true whenever  $(i = 0) \wedge (j = 0)$  is true, T2 has to be enabled whenever T3 is enabled; if we always choose to execute T2 whenever T3 is enabled, we'll never reach a state  $x = true$ .

## Problem 7

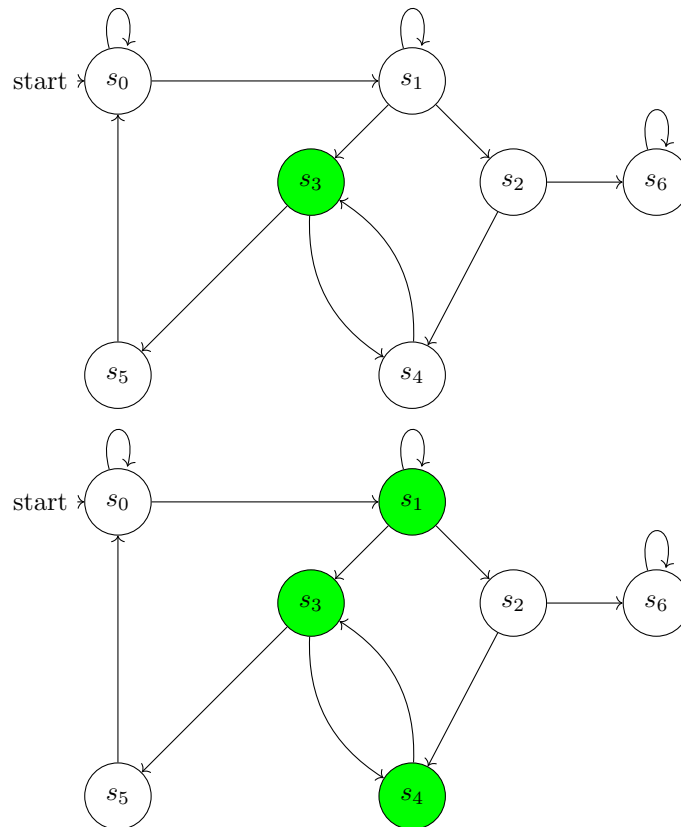
(a)

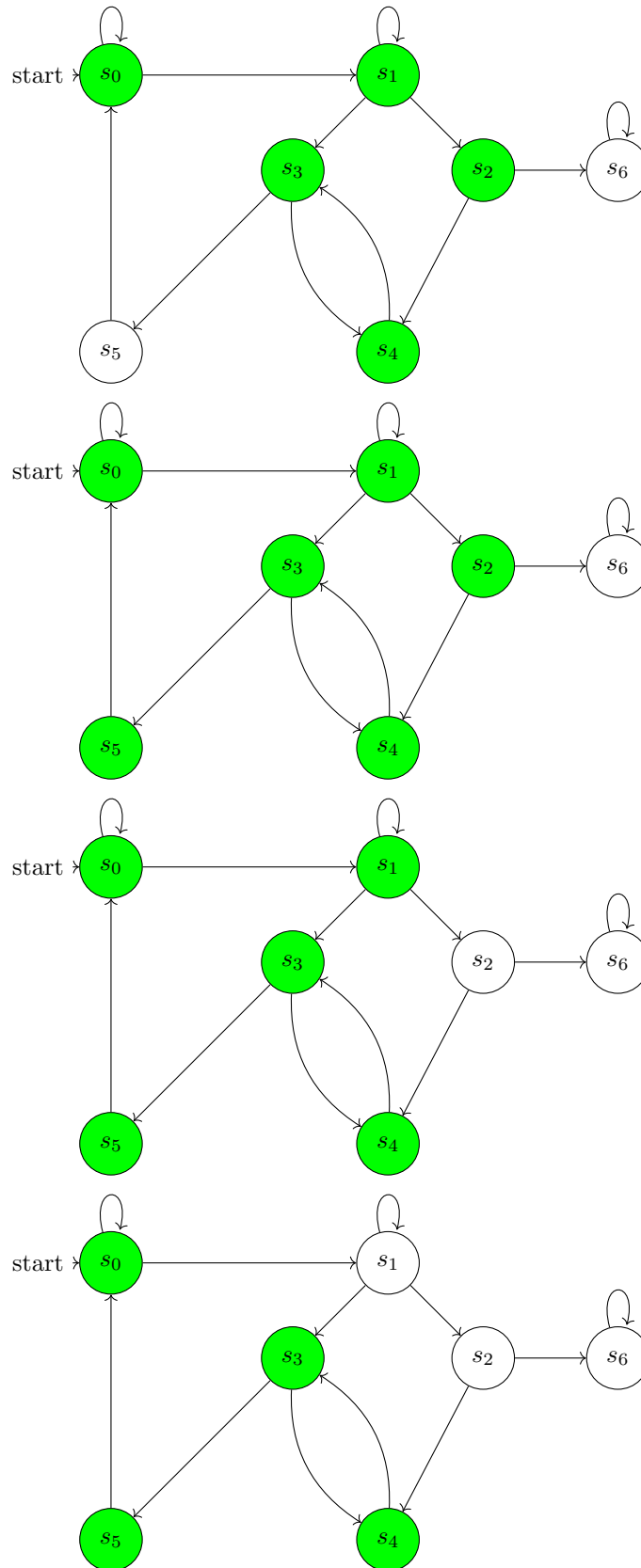


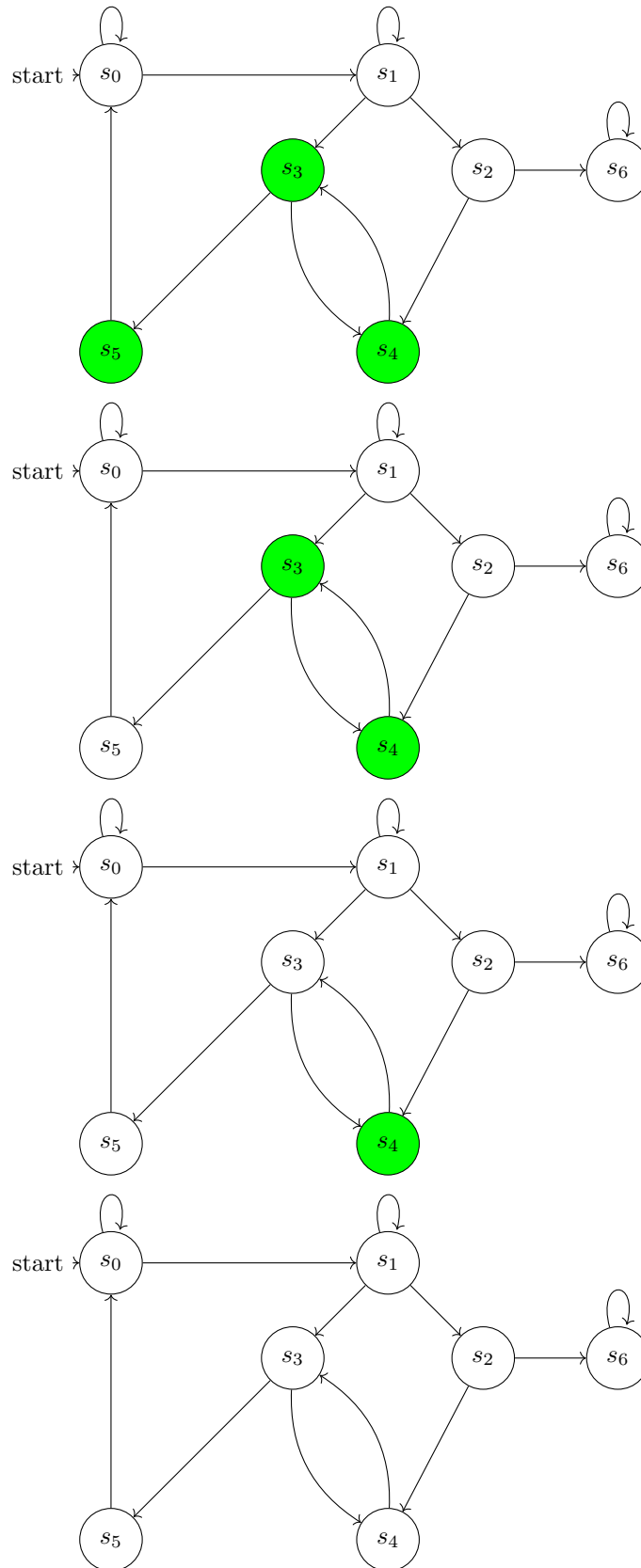


No state with unlabeled successor, so cannot remove any state. Satisfies **AGEF***green*.

(b)







No labeled state after running fixpoint; doesn't satisfy **AGEF***green*.