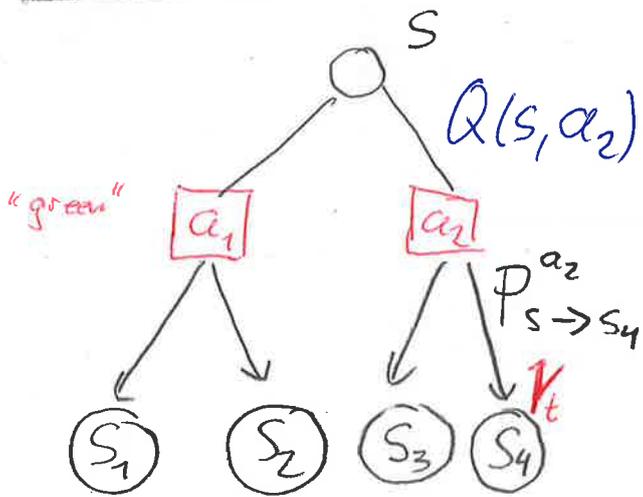


Blackboard RL1 : Q-values



"branching ratio"

Transition probability

$$P_{s \rightarrow s_4}^{a_2} = P(s' = s_4 | a_2, s)$$

↑
next state

- actual reward at time t : R_t
- expected reward for this "branch"

$$R_{s \rightarrow s_4}^{a_2} = E(R_t | s' = s_4, a_2, s)$$

↑
reward received

↑
end up in s_4

↑
take a_2

↑
start in s

- expected reward for action a_2

$$Q(s, a_2) = E(R_t | a_2, s)$$
$$= \sum_{s'} P_{s \rightarrow s'}^{a_2} \cdot R_{s \rightarrow s'}^{a_2}$$

↑
all possible
"next states"

Blackboard RL1-2 = Exercise 1

(2)

$Q = \text{expected reward} \approx \text{empirical mean } r. = \hat{Q}$

$\hat{Q}^{(k-1)}(s, \alpha)$ after $k-1$ trials (playing action α)

$$\hat{Q}^{(k-1)}(s, \alpha) = \frac{1}{k-1} (\underbrace{r_1 + r_2 + \dots + r_{k-1}}_{\substack{\uparrow \\ \text{2nd time action } \alpha}})$$

after k trials

$$\begin{aligned} \hat{Q}^{(k)}(s, \alpha) &= \frac{1}{k} (\underbrace{r_1 + r_2 + \dots + r_{k-1}}_{\text{from previous step}} + r_k) \\ &= \frac{k-1}{k} \cdot \hat{Q}^{(k-1)}(s, \alpha) + \frac{1}{k} r_k \\ &= \cancel{\frac{k}{k}} \hat{Q}^{(k-1)}(s, \alpha) + \frac{1}{k} r_k - \frac{1}{k} \hat{Q}^{(k-1)}(s, \alpha) \end{aligned}$$

$$\text{Eq. (1)} \quad \Delta \hat{Q}(s, \alpha) = \hat{Q}^{(k)}(s, \alpha) - \hat{Q}^{(k-1)}(s, \alpha) = \frac{1}{k} [r_k - \hat{Q}^{(k-1)}]$$

$$\Rightarrow \boxed{\eta = \frac{1}{k}}$$

Theorem (i): if $E[\Delta \hat{Q}(s, \alpha)] = 0$ (H)

then $E[\hat{Q}(s, \alpha)] = \sum_{s'} P_{s \rightarrow s'}^\alpha R_{s \rightarrow s'}^\alpha$
 ↑
 expectation

proof: (H) Eq. (1) of slide

$$E[\Delta \hat{Q}(s, \alpha)] \stackrel{\downarrow}{=} 0 = E[r_t - \hat{Q}(s, \alpha)]$$

↑
fluctuates around zero

$$0 = E[r_t] - E[\hat{Q}(s, \alpha)]$$

$$0 = \sum_{s'} P_{s \rightarrow s'}^\alpha R_{s \rightarrow s'}^\alpha - E[\hat{Q}(s, \alpha)]$$

(ii) Fluctuations: role of η is qualitatively obvious.
 smaller $\eta \Rightarrow$ smaller fluctuation

↑
 \hat{Q} fluctuates around $E[Q(s, \alpha)] = Q(s, \alpha)$
 ↑
 expectation

Blackboard RL1-4: Exercise 2

(4)

update with $\Delta Q(s, a) = 0.2 \cdot [r_t - Q(s, a)]$ (*)

2.1. initialise $Q(s, a_1) = Q(s, a_2) = 0$

$t=1$, action a_1 ; $r_t = 1 \Rightarrow \underline{Q(s, a_1) = 0.2}$

$t=2$, action a_2 ; $r_t = 0.4 \Rightarrow \boxed{Q(s, a_2) = 0.08}$

2.2. $t=3$, best action = a_1 ; $r_t = 0$

$Q(s, a_1) \leftarrow Q(s, a_1) + 0.2[0 - 0.2]$; $\underline{Q(s, a_1) = 0.16}$

$t=4$, best action a_1 ; $r_t = 0$

$Q(s, a_1) \leftarrow Q(s, a_1) + 0.2[0 - 0.16]$
 $0.16 - 0.032$ $\underline{Q(s, a_1) = 0.128}$

$t=5$, best action a_1 ; $r_t = 0$

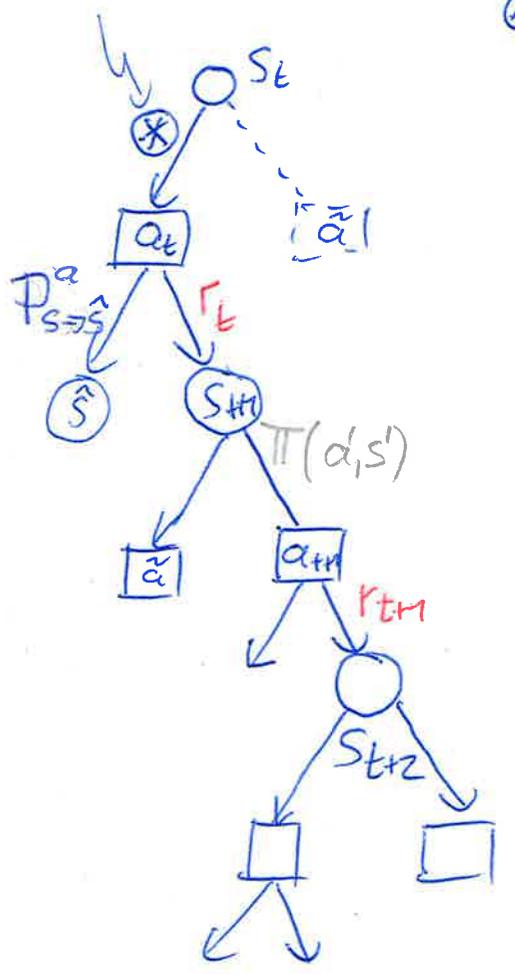
$Q(s, a_1) \leftarrow 0.128 - 0.2 \cdot 0.128$; $\underline{Q(s, a_1) \approx 0.102}$

$\Rightarrow \underline{a_1 \text{ remains "best action" for several steps!}}$

2.3 actual values

$\left. \begin{array}{l} Q(s, a_1) = 0.25 \\ Q(s, a_2) = 0.30 \end{array} \right\} \Rightarrow \underline{a_2 \text{ is best action}}$

we start here



⊗ total reward collected in single trial starting in \$S\$ with action \$a_t\$

$$\begin{aligned}
 R^{tot}(s_t, a_t) &= r_t + \gamma r_{t+1} + \gamma^2 r_{t+2} + \gamma^3 r_{t+3} + \dots \\
 &= r_t + \gamma [r_{t+1} + \gamma r_{t+2} + \gamma^2 r_{t+3} + \dots] \\
 &= r_t + \gamma \cdot R^{tot}(s_{t+1}, a_{t+1})
 \end{aligned}$$

total reward (single trial) starting from \$s' = s_{t+1}\$ with \$a_{t+1}\$

now we look at diagram to calculate expectation

$$\begin{aligned}
 E(R^{tot}(s_t, a_t)) &= E(r_t + \gamma R^{tot}(s_{t+1}, a_{t+1})) \\
 &= \sum_{s'} P_{S \to s'}^{a_t} [R_{S \to s'}^{a_t} + \gamma E(R^{tot}(s'))] \\
 &= \sum_{s'} P_{S \to s'}^{a_t} [R_{S \to s'}^{a_t} + \gamma \cdot \sum_{a'} \Pi(a', s') E(R(s', a'))] \\
 Q(s_t, a_t) &= \sum_{s'} P_{S \to s'}^{a_t} [R_{S \to s'}^{a_t} + \gamma \sum_{a'} \Pi(a', s') Q^{tot}(s', a')]
 \end{aligned}$$

Blackboard RL3-6 - SARSA

(6)

from diagram

$$Q(s, a) \approx r_t + \underset{\substack{\text{discount} \\ \downarrow}}{\gamma} \cdot Q(s', a')$$

$$0 \approx r_t + \gamma \cdot Q(s', a') - Q(s, a)$$

proposed update

$$\Delta Q(s, a) = \gamma [r_t + \gamma \cdot Q(s', a') - Q(s, a)]$$

check:

$$\text{if } r_t > \underbrace{\gamma \cdot Q(s', a') - Q(s, a)}_{\substack{\text{expected reward} \\ \text{for this transition}}} \Rightarrow \text{increase } Q(s, a)$$

↑
actual
reward

Blackboard 7 - Convergence / Blackboard 7

SARSA update

$$\Delta Q(s, a) = \gamma [r_t + \gamma Q(s', a') - Q(s, a)]$$

hypothesis

$$E[\Delta Q(s, a)] \stackrel{!}{=} 0 = E \left[r_t + \gamma Q(s', a') - Q(s, a) \right]$$

↑
starting in s_t with a

$$0 = \sum_{s'} P_{s \rightarrow s'}^a \left[R_{s \rightarrow s'}^a + \gamma \sum_{a'} \pi(s', a') Q(s', a') \right] - Q(s, a)$$

\Rightarrow Bellman \checkmark

in order to evaluate expectations:

- look at graph!
- if I am in s , all remaining expectations are "given s "
- if I am in a branch (s, a)
all remaining expectations are given s and a