Tiger Problem

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State space (tiger lies left, tiger lies right)

$$\{S\} = \{s_0, s_1\} \tag{1}$$

Belief state

$$b = (x, 1 - x) \tag{2}$$

Action (listen left; open left, open right)

$$\{a\} = \{a_0; a_1, a_2\} \tag{3}$$

Observation (tiger noise, no tiger noise)

$$\{z\} = \{z_0, z_1\} \tag{4}$$

Belief state transition (z observed)

$$b'(s')\Big|_{a,b,z} = \frac{1}{p(z|b,a)} \mathbb{P}[z|s',a] \sum_{s} \mathbb{P}[s'|s,a]b(s) \equiv b^{a,z}$$

$$\tag{5}$$

Belief state transition (z not observed)

$$\mathbb{P}[b'\big|b,a] = \begin{cases} \sum_{z} \mathbb{P}[z\big|b,a] = \sum_{z} \sum_{s} \mathbb{P}[z\big|s,a]b(s) , & \text{for } \forall z \text{ s.t. } b' = b^{a,z} \\ 0 , & \text{if no } z \text{ satisfies } b' = b^{a,z} \end{cases}$$

$$\tag{6}$$

Value iteration (k indicates value iteration number)

$$V^{k+1}(b) = \max_{a} \left\{ \sum_{s} r(a,s)b(s) + \gamma \sum_{z} \max_{\mathbf{j}} \left(\sum_{s'} \alpha_{\mathbf{j}}^{k-1}(s') \mathbb{P}\left[z\big|s',a\right] \sum_{s} \mathbb{P}\left[s'\big|s,a\right] b(s) \right) \right\} \tag{7}$$

 α -vector pool: $\Gamma^k \equiv \{(\alpha_1, a_{l_1})^k, \cdots, (\alpha_{\mathbf{j}}, a_{l_{\mathbf{j}}})^k, \cdots \}$ Given b, α -vector backup:

1. For $\forall a, \forall z, \text{ find}$

$$\alpha^{k-1,*}\big|_{a,z} = \arg\max_{\alpha_{\mathbf{j}}^{k-1} \in \Gamma^{k-1}} \sum_{s'} \alpha_{\mathbf{j}}^{k-1}(s') \mathbb{P}\left[z\big|s',a\right] \sum_{s} \mathbb{P}\left[s'\big|s,a\right] b(s) \tag{8}$$

2. For $\forall a, \forall s, \text{ find}$

$$\alpha^{k}(s)\big|_{a} = r(a,s) + \gamma \sum_{z} \sum_{s'} \alpha^{k-1,*}(s')\big|_{a,z} \mathbb{P}\left[z\big|s',a\right] \mathbb{P}\left[s'\big|s,a\right] \tag{9}$$

3. Find

$$a^* = \arg\max_{a} \sum_{s} \alpha^k(s) |_a b(s)$$
 (10)

4. [optional] Append α -vector pool (with redundancy)

$$\bar{\Gamma}^k \leftarrow \left\{ \Gamma^{k-1}, (\alpha^k|_{a^*}, a^*) \right\} \tag{11}$$

Definition 1. Dominate subset of vectors:

Let $\bar{\Gamma} = \{\alpha_1, \dots, \alpha_q\}$ be a set of vectors in \mathbb{R}^n , b be a vector in \mathbb{R}^n . Γ is called the dominate subset of $\bar{\Gamma}$, iff for $\forall \alpha_i \in \Gamma \subseteq \bar{\Gamma}$, $\exists b$, s.t.

$$\alpha_i^T b > \alpha_{\{1,\cdots,q\}\setminus i}^T b \tag{12}$$

subject to

$$I_{n \times n} b \ge 0$$

$$(1, \dots, 1) b = 1$$

$$(13)$$

 Γ can be obtained from $\bar{\Gamma}$ by solving a linear programming problem by e.g. simplex method. To implement point-based value iteration on all $\mathcal{R}^L(b_0)$ (\mathcal{R} indicates reachable belief states, L indicates maximum depth)

- 1. **Initialize** Belief tree $T(b_0)$ with maximum depth L, consisted of action nodes and belief nodes.
- 2. repeat
- 3. Choose all belief nodes b_l from $T(b_0)$
- 4. For each b_l , backup $\bar{\Gamma}$
- 5. $\Gamma \leftarrow \bar{\Gamma}$
- 6. **until** the max increament of $V(b_l)$ is smaller than ϵ
- 7. return Γ

Instead of the above approach, consider Perseus algorithm

Perseus Algorithm

- 1. initialize
- 2. Initialize Belief tree $T(b_0)$ with maximum depth L
- 3. $B \leftarrow \text{random subset of non-leaf belief nodes of } T(b_0)$
- 4. Lower bound of α -vectors

$$\alpha \leftarrow \frac{\min\{r\}}{1 - \gamma}(1, 1) \tag{14}$$

- 5. Iteration count $k \leftarrow 0$
- 6. α -vector pool $\Gamma^0 \leftarrow \{\alpha\}$
- 7. repeat
- 8. for b in B:
- 9. Store $\alpha^k(b) \leftarrow \arg\max_{\alpha \in \Gamma^k} \langle \alpha, b \rangle$
- 10. Store $V^k(b) \leftarrow \langle \alpha^k(b), b \rangle$
- 11. $\Gamma^{k+1} \leftarrow \varnothing$
- 12. Not-improved belief points $\tilde{B} \leftarrow B$

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while \tilde{B} is not \varnothing:
13.
                                b \,\leftarrowa random sample from \tilde{B}
14.
                                \alpha \, \leftarrow \, \mathrm{backup}(b,\Gamma^k)
15.
                                if \langle \alpha, b \rangle < V^k(b):
16.
                                           \alpha \leftarrow \alpha^k(b)
17.
                                for b' in \tilde{B}:
18.
                                           if \langle \alpha, b' \rangle \geq V^k(b'):
19.
                                                     \tilde{B} \leftarrow \tilde{B} \setminus b'
20.
                                \Gamma^{k+1} \leftarrow \Gamma^{k+1} \cup \alpha
21.
22.
                    k \leftarrow k+1
23. until convergence
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1 Continuous poMDP

Bellman equation

$$V(b) = \sup_{a \in A} \left\{ \langle r_a, b \rangle + \gamma \int_z \mathbb{P}[z|b, a] V(b^{a,z}) \, \mathrm{d}z \right\}$$
 (15)