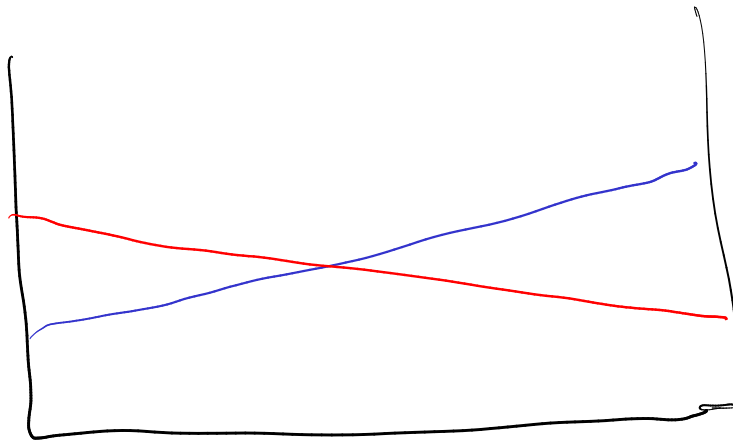
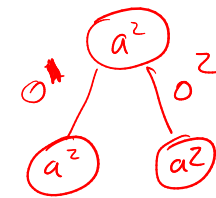
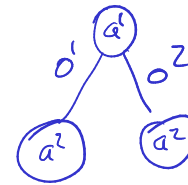


Offline POMDP Algorithms



$b(\cdot)$



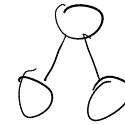
Last time: POMDP Value Iteration (horizon d)

$$\Gamma^0 \leftarrow \emptyset$$

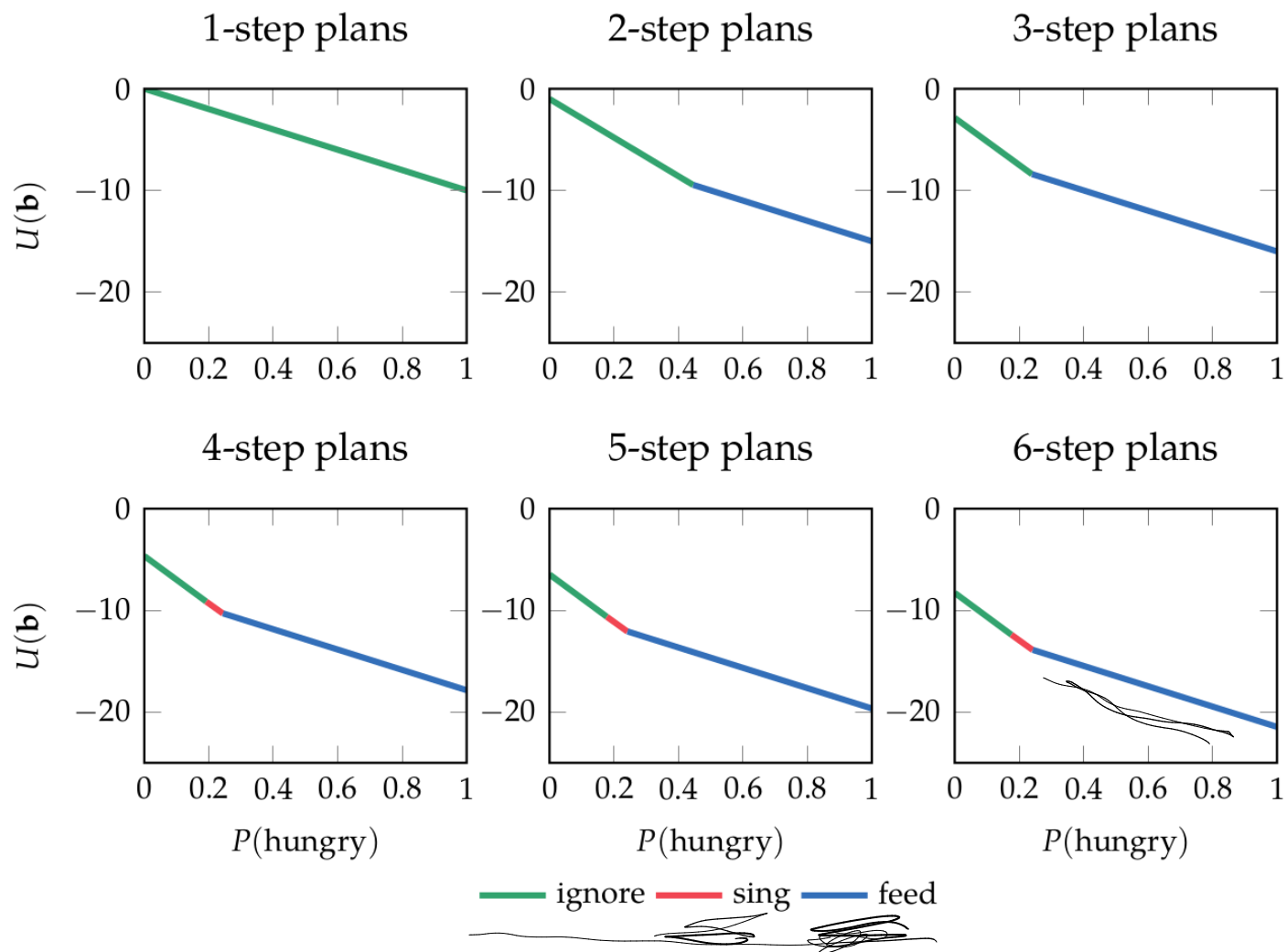
for $n \in 1 \dots d$

Construct Γ^n by expanding with Γ^{n-1}

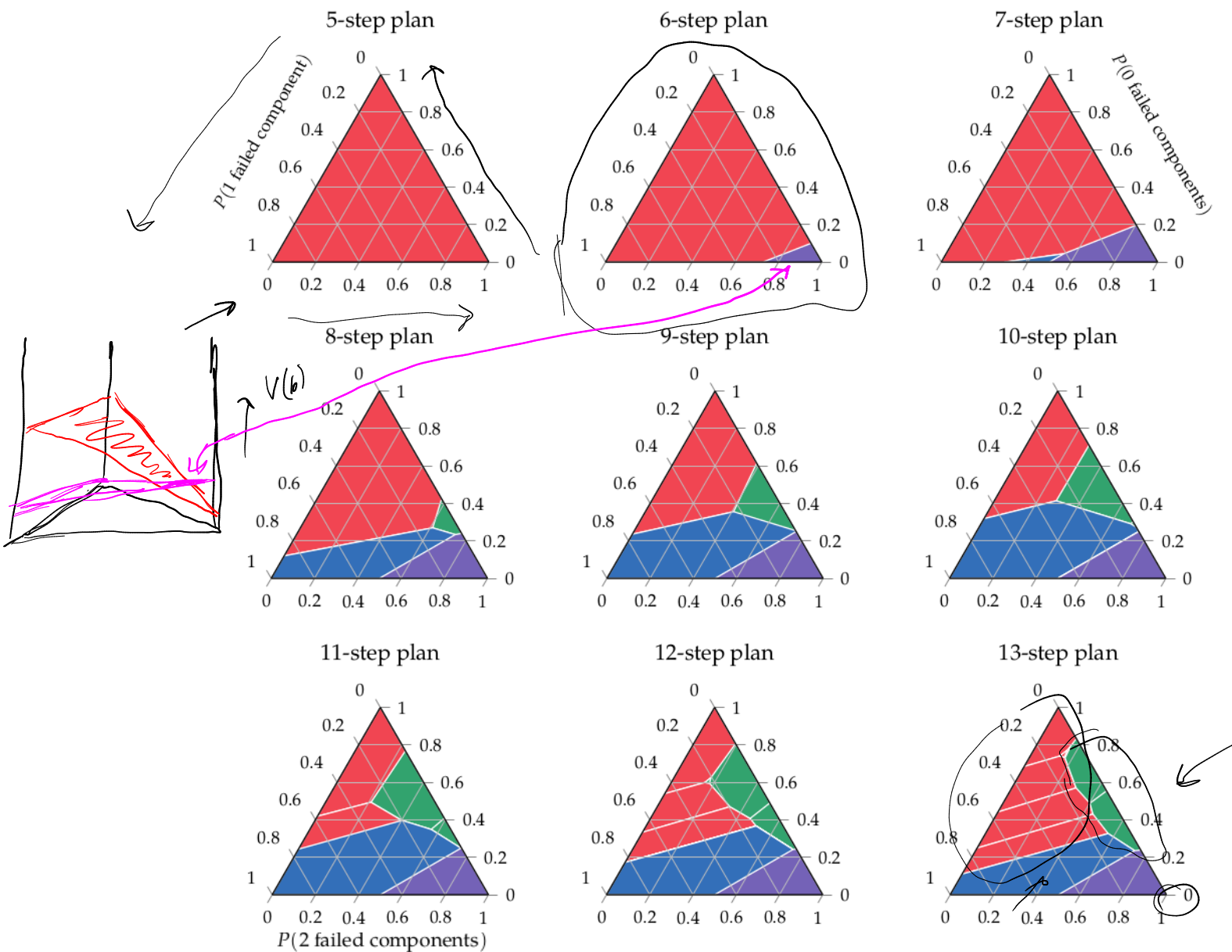
Prune Γ^n



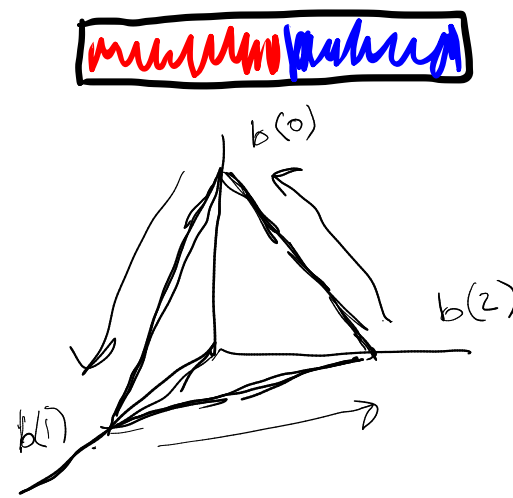
Finite Horizon POMDP Value Iteration



Finite Horizon POMDP Value Iteration



$$B = \Delta(s)$$



Infinite-Horizon POMDP Lower Bound Improvement

$$\frac{R}{1-\gamma}$$

Infinite-Horizon POMDP Lower Bound Improvement

$\alpha_a = (I - \gamma T^a)^{-1} R^a$
 ↙ always execute same action
 $\Gamma \leftarrow$ blind lower bound

loop

$\Gamma \leftarrow \Gamma \cup \text{backup}(\Gamma)$

$\Gamma \leftarrow \text{prune}(\Gamma)$

A survey of point based POMDP solvers

backup

$$\Gamma' = \bigcup_{a \in A} \Gamma^a$$

$$\Gamma^a = \bigoplus_{o \in O} \Gamma^{a,o}$$

$$\Gamma^{a,o} = \left\{ \frac{1}{|O|} r_a + \alpha^{a,o} : \alpha \in \Gamma \right\}$$

$$\alpha^{a,o}[s] = \sum_{s'} \mathbb{P}(o|a,s) T(s'|s,a) \alpha[s']$$

$$O(|\Gamma| |A| |O| |S|^2 + |A| |S| |\Gamma|^{10A})$$

$$\Gamma' \oplus \Gamma^2 = \{ \alpha_1 + \alpha_2 : \alpha_1 \in \Gamma', \alpha_2 \in \Gamma^2 \}$$

Point-Based Value Iteration (PBVI)

point-backup(Γ, b)

for $a \in A$

for $o \in O$

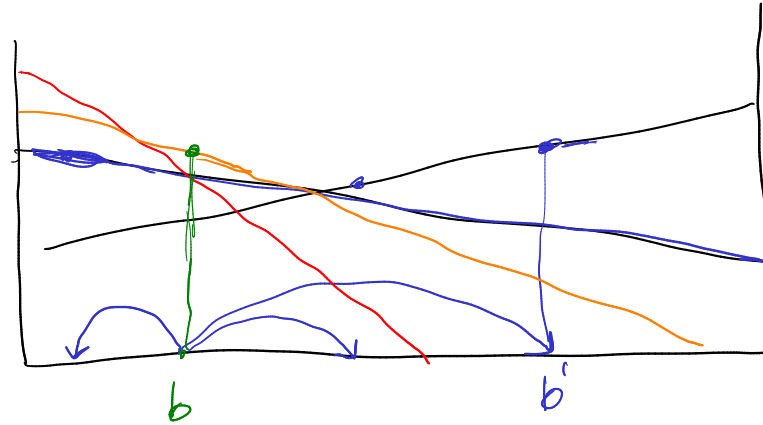
$$\underline{b'} \leftarrow \underline{\tau(b, a, o)}$$

$$\underline{\alpha_{a,o} \leftarrow \operatorname{argmax}_{\alpha \in \Gamma} \alpha^\top b'}$$

for $s \in S$

$$\alpha_a[s] = R(s, a) + \gamma \sum_{s', o} T(s' | s, a) Z(o' | a, s') \alpha_{a,o}[s']$$

return $\operatorname{argmax}_{\alpha_a} \alpha_a^\top b$



B

If we perform a backup
for each $b \in B$

$$O(|A||O||\Gamma||S|^2 + |B||A||S||O|)$$

Original PBVI

how do we choose B

$$B \leftarrow b_0$$

loop

for $b \in B$

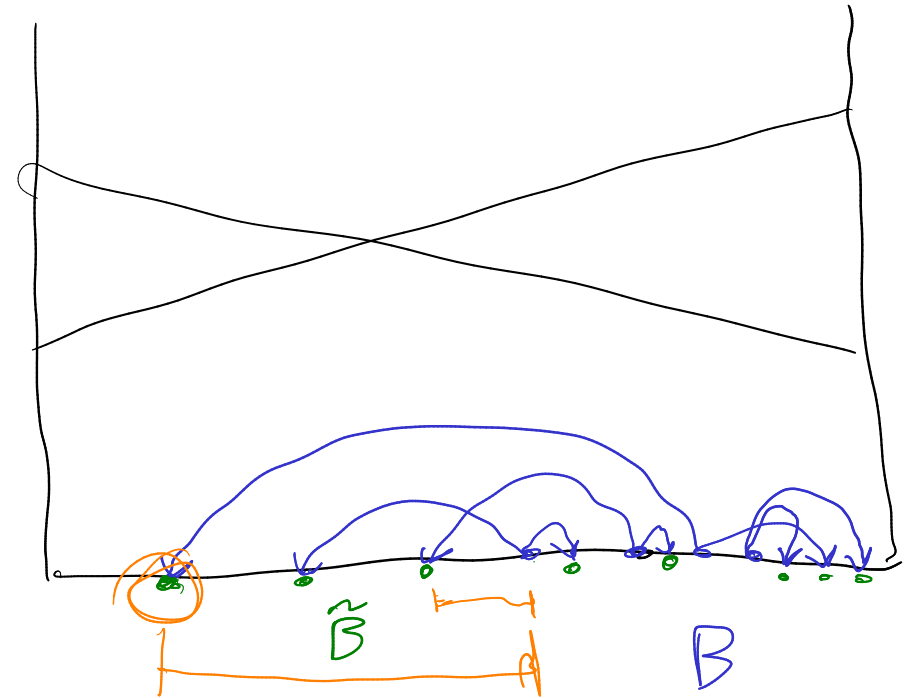
$$\Gamma \leftarrow \Gamma \cup \{\text{point_backup}(\Gamma, b)\}$$

for $b \in B$

$$\tilde{B} \leftarrow \{\tau(b, a, o) : a \in A, o \in O\}$$

$$B' \leftarrow B' \cup \left\{ \underset{b' \in \tilde{B}}{\operatorname{argmax}} \|B, b'\| \right\}$$

$$B \leftarrow B \cup B'$$



PERSEUS: Randomly Selected Beliefs

Two Phases:

1. Random Exploration
2. Value Backup

Random Exploration:

$$B \leftarrow \emptyset$$

$$b \leftarrow b_0$$

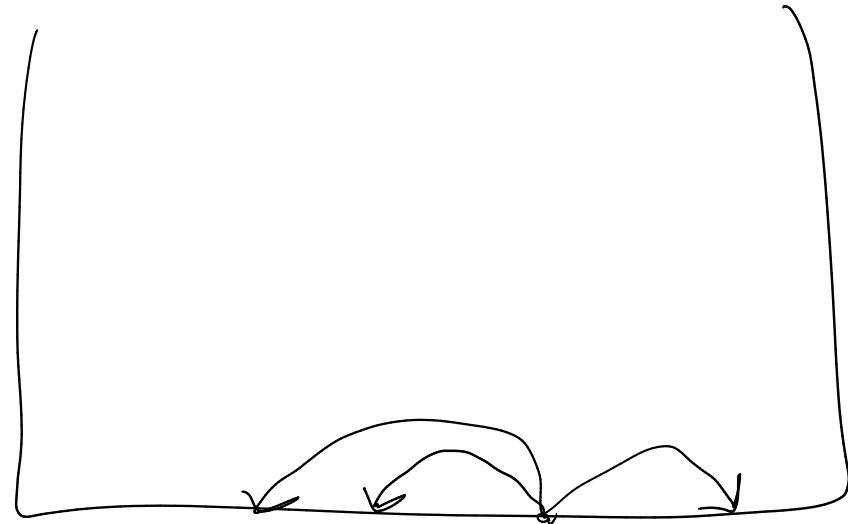
loop until $|B| = \underline{n}$

$$a \leftarrow \text{rand}(A)$$

$$o \leftarrow \text{rand}(P(o \mid b, a))$$

$$b \leftarrow \tau(b, a, o)$$

$$\underline{B} = B \cup \{b\}$$



Heuristic Search Value Iteration (HSVI)

while $\bar{V}(b_0) - \underline{V}(b_0) > \epsilon$
 explore($b_0, 0$)

$\bar{V}(b)$
 upper bound
 for $b \in \mathcal{B}$

$\underline{V}(b)$
 lower bound

function explore(b, t)

if $\bar{V}(b) - \underline{V}(b) > \epsilon \gamma^t$

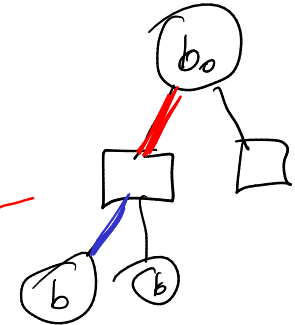
$a^* = \operatorname{argmax}_a \bar{Q}(b, a)$

$o^* = \operatorname{argmax}_o P(o \mid b, a) (\bar{V}(\tau(b, a^*, o)) - \underline{V}(\tau(b, a^*, o)) - \epsilon \gamma^t)$

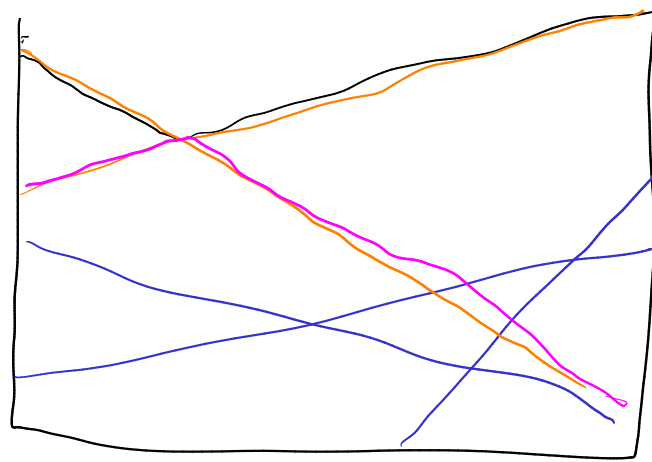
explore($\tau(b, a^*, o^*), t + 1$) ←

$\underline{\Gamma} \leftarrow \underline{\Gamma} \cup \text{point_backup}(\underline{\Gamma}, b)$ ←

$\bar{V}(b) = \underline{B}_b [\bar{V}(b)]$ ←



Sawtooth Upper Bounds



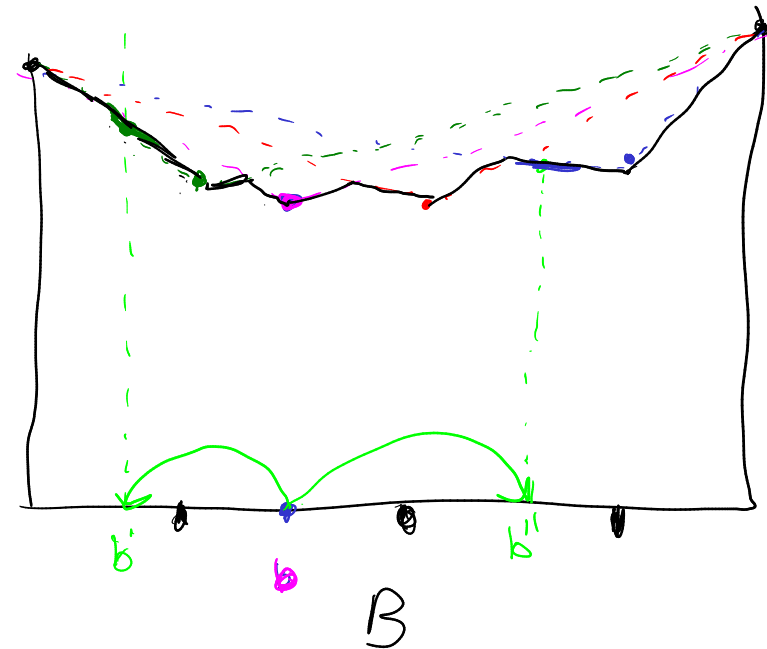
~~$\min_a \alpha^T b$~~

lower bound

$$\underline{V}(b) = \max_{\alpha \in \Gamma} \alpha^T b$$

$$\underline{B}_b[\underline{V}](b) = \max_a R(b, a) + \gamma \sum_{\rho} P(o|b, a) \underline{V}(\tau(b, a, \rho))$$

\underline{b}'



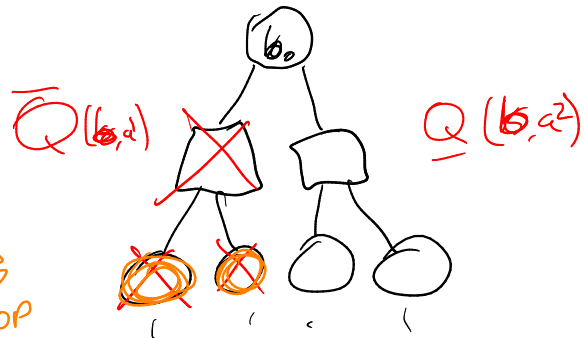
SARSOP

Successive Approximation of Reachable Space under Optimal Policies

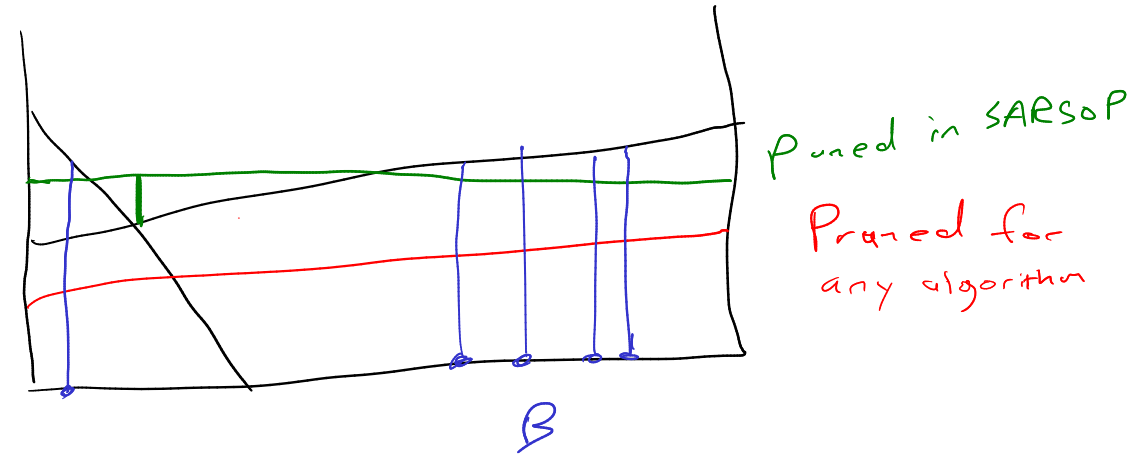
Similar to HSVI

HSVI
 $B \subset R$
 \uparrow
 reachable

SARSOP
 $B \subset R^*$
 \uparrow
 reachable under
 optimal policy



if $\bar{Q}(b, a') < \underline{Q}(b, a'')$
 then prune all b
 below (b, a')



Witness (α-vector value iteration) : ~ 20 states

SARSOP : 10,000 - 100,000 states

Offline POMDP Algorithms

Policy Graphs

Monte Carlo Value Iteration (MCVI)