

Last Time
 α vectors

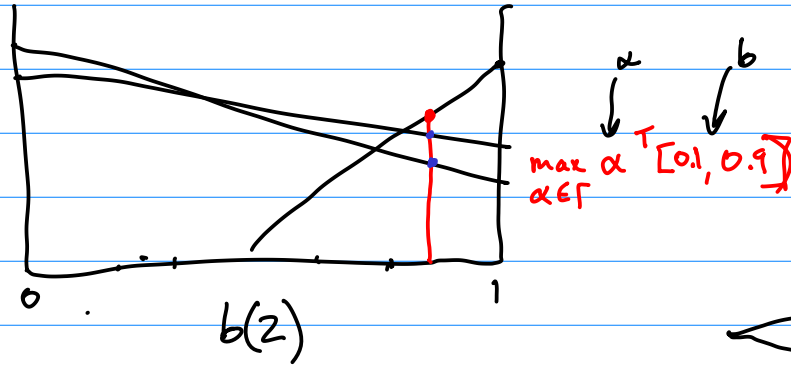
$\alpha[s]$

$\downarrow \downarrow \downarrow$
 $[0.4, 0.3, 0.2]$

set Γ

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

$S = \{1, 2\}$



Today

Offline POMDP

"Survey of
 Point-Based
 POMDP Solvers"

Name	S	A	O
Exact VI	D (10)	D	D
PBVI	D	D	D
Perseus	D	D	D
HSVI	D	D	D
SARSOP	D (1000)	D	D
MCVI	C	D	D/C
LQG	C	C	C

bigger

(Linear
 Gaussian
 Quadratic)

Exact VI

loop

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

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

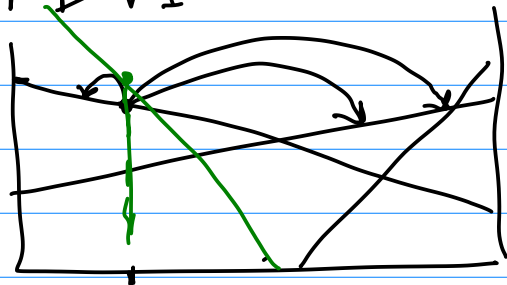
$$\Gamma^{a,o} = \left\{ \frac{1}{|O|} r_a + \alpha^{a,o} : \alpha \in \Gamma \right\}$$

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

for 1 iteration

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

PBVI



→ Backup Belief(Γ, b)

for $a \in A$

for $o \in O$

$$b' = \tau(b, a, o)$$

$$\alpha_{a,o} = \argmax_{\alpha \in \Gamma} \alpha^T b'$$

for $s \in S$

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

return $\argmax_{\alpha} \alpha^T b$

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

How we choose B

set of Beliefs

Original PBVI

$$B = \{b_0\}$$

loop $B' = \emptyset$

for $b \in B$

$$\Gamma = \Gamma \cup \text{Backup Belief}(\Gamma, b)$$

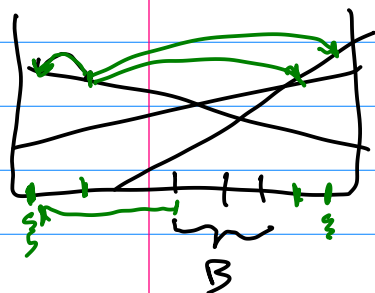
for $b \in B$

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

$$B' = B' \cup \argmax_{b \in \tilde{B}} \|B, b'\|$$

← add furthest b away from B

$$B = B \cup B'$$



$$B = \{b^1, b^2, b^3\}$$

$$B' = \{b^4, b^5\}$$

$$B \cup B' = \{b^1, \dots, b^5\}$$

Perseus - randomly choose B

$\rightarrow B = \emptyset$
 $\rightarrow b = b_0$
 loop until $|B| = n$
 $\quad a = \text{rand}(A)$
 $\quad o = \text{rand}(P(o|b, a))$
 $\quad B = B \cup \{\tau(b, a, o)\}$

HSV I Heuristic Search Value Iteration

$\bar{V}(b)$ upper bound

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

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

$\text{explore}(b, t)$

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

$a^* = \arg\max_a \bar{Q}(b, a)$

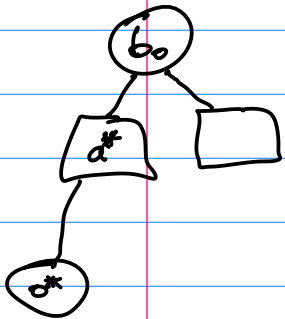
$o^* = \arg\max_o (P(o|b, a^*) (\bar{V}(\tau(b, a^*, o))$

$- \underline{V}(\tau(b, a^*, o)) - \epsilon \gamma^{t+1})$

$\text{explore}(\tau(b, a^*, o^*), t+1)$

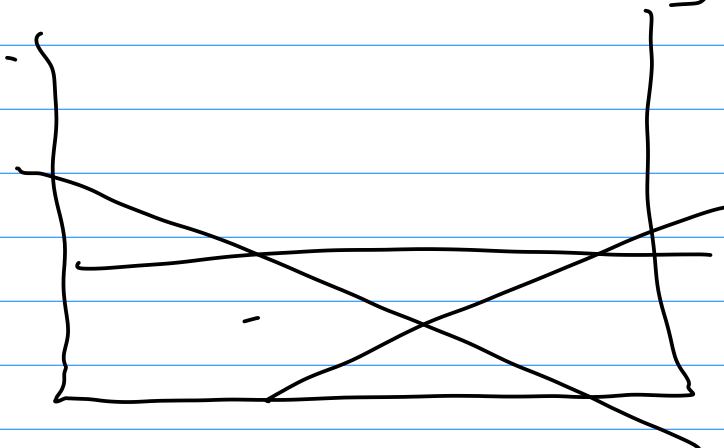
$\underline{\Gamma} = \underline{\Gamma} \cup \text{Backup Belief}(\underline{\Gamma}, b)$

$\bar{V}(b) = B_b[\bar{V}(b)]$



\rightarrow

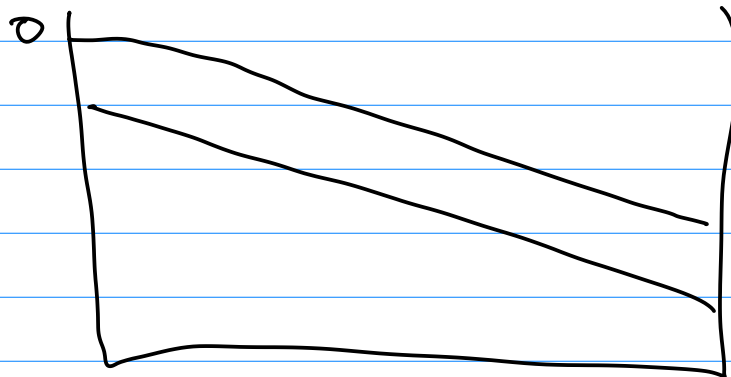
\rightarrow



Baby POMDP

(+)

(-)

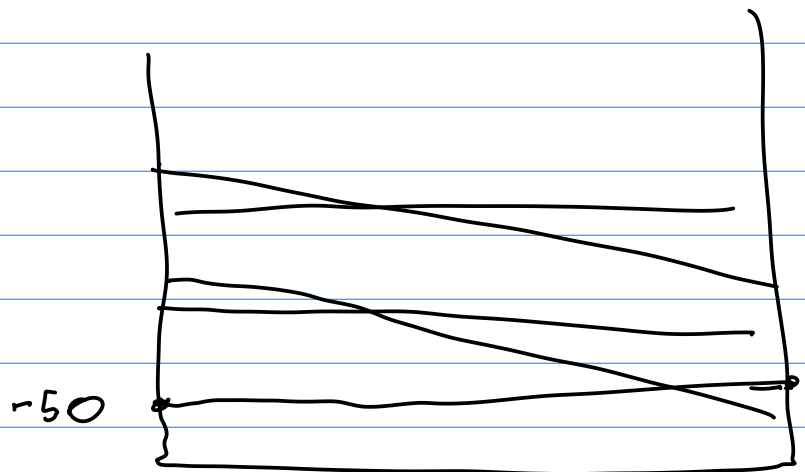


$$\frac{R}{1-\gamma}$$

Open-Loop Policy

Never Feed

$$\frac{-5}{1-0.9} = -50$$



Upper Bounds

Sawtooth Upper bound

Finding points to interpolate between is an LP



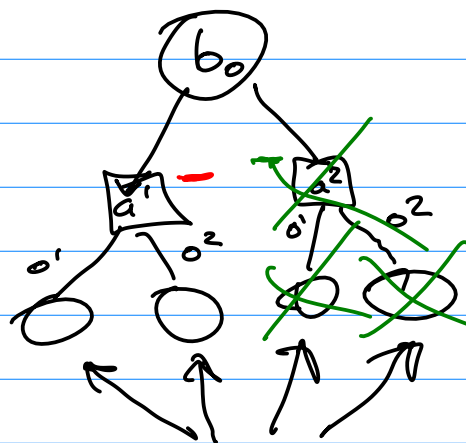
$$B_b[\bar{V}(b)] = \max_a R(b,a) + \gamma \sum_o P(o|b,a) \bar{V}(\tau(b,o))$$

SARSAOP

Successive Approximation of
Reachable Space under Optimal
Policies

Similar to HSUI

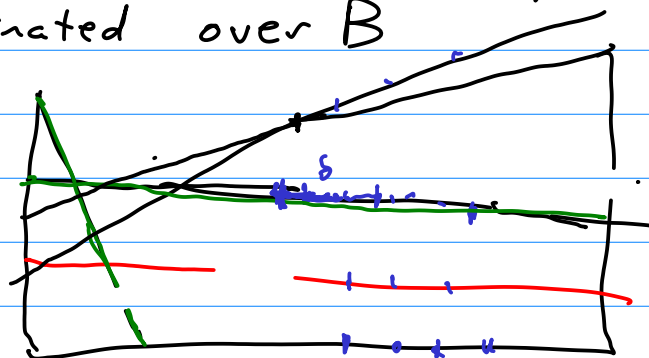
BCR BCR*



Belief points that could
be in B

if $\bar{Q}(b,a^1) < \underline{Q}(b,a^2)$
then prune all b below (b,a^2)
from B

Instead of pruning α that are dominated
over the whole belief space, prune α
dominated over B



Pruned by Any

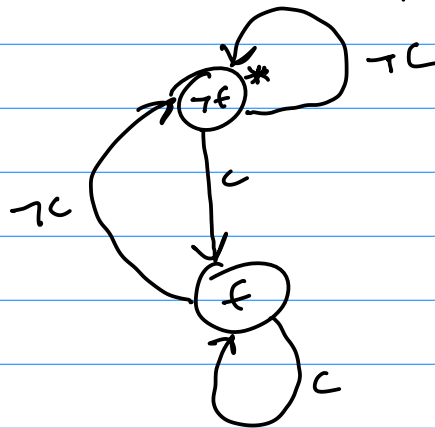
Pruned by
SARSAOP

B

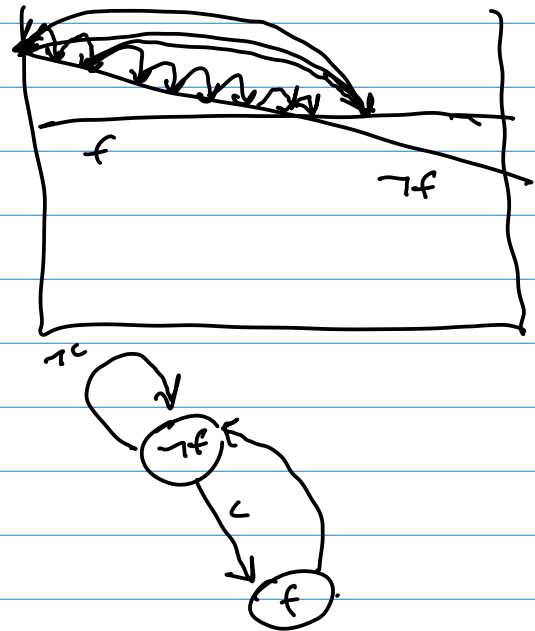
Policy Graph

vertex labeled with a
edge labeled with 0

Feed when Crying



Policy Graph \rightarrow α vector
evaluate plan
 α vectors \rightarrow policy graphs



MCVI "Monte Carlo Value Iteration"

$$V_G(b) = \max_{\alpha \in \mathcal{A}} \int \alpha_v(s) b(s) ds \quad \leftarrow \text{Approx w/MC}$$

Fig.

v

MC-Backup (G, b, N)

$$R_a = 0 \quad V_{a,o,v} = 0$$

for $a \in A$

for i in $1:N$

$s_i \leftarrow \text{sample}(b)$

$s'_i, o_i, r_i \leftarrow G(s_i, a)$

$R_a + r_i$

for $v \in G$

$V_{a,o_i,v} = V_{a,o_i,v} + \text{Simulate}(G, v, s'_i, L)$

for o in O

$$V_{a,o} = \max_{v \in G} V_{a,o,v}$$

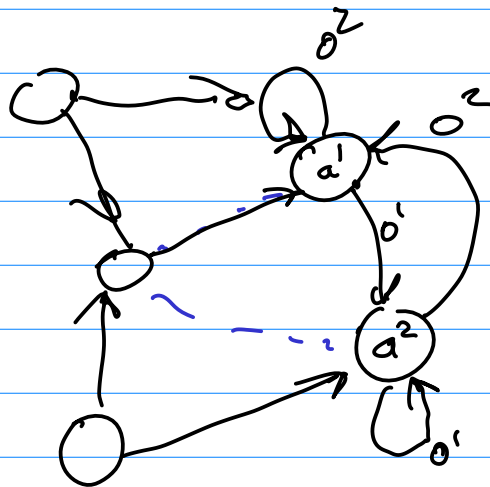
$$v_{a,o} = \text{argmax}_{v \in G} V_{a,o,v}$$

$$V_a = R_a + \gamma \sum_o V_{a,o} / N$$

$$V^* = \max_a V_a$$

$$a^* = \text{argmax}_a V_a$$

add new node to G labeled with a^*



V = value

v = node in graph