

Last-Time
Discrete Bayesian Filter

This Time

POMDPS

Particle Filters

Value of Information

Outcome Allectory MDP known model
Model Epistemic Static RL

State Epistemic Dynamic POMDP

MDP (S, A,T, R, Y) POMDP (S, A,T,R, D, Z,Y)

Examples -Automated Driving -ACASX - Space Exploration - Disease Monitoring/Treatment - HVAC - Water Management ~ Wildfire - H RI

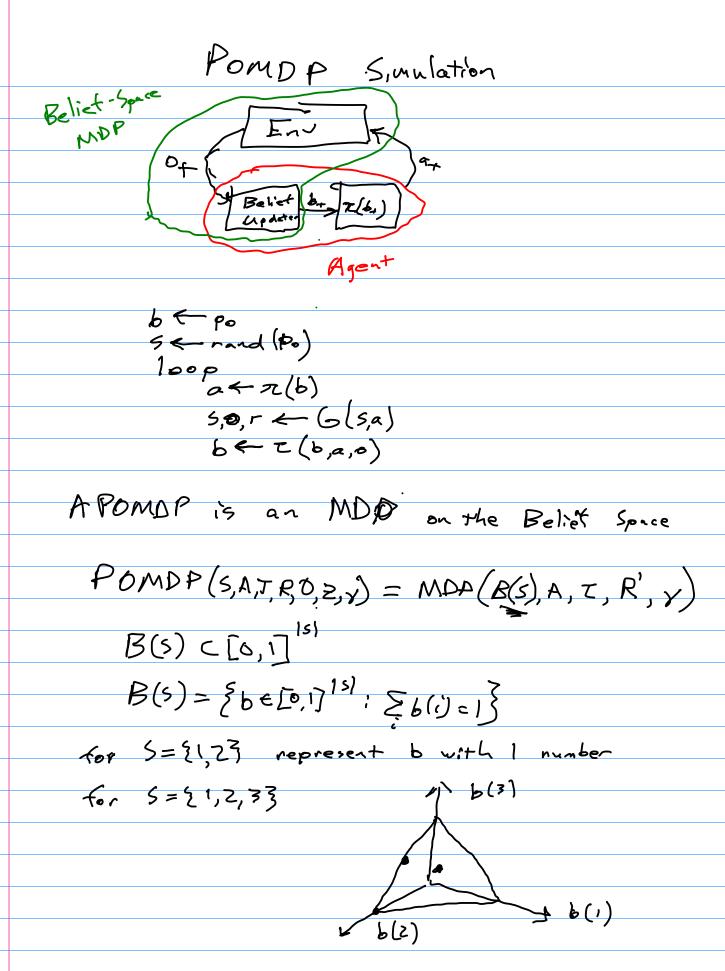
50 far
$$O_{+}=S_{+}$$
 $\pi(S_{+})$
now $O_{+}\sim Z(a_{+-1},S_{+})$

MDPs: optimal policy based on last state

POMDPs! optimal policy based on last obx

- history: h= (po, ao, 0, 1 a+-1,0+)

- belief: b+(s) = P(s+=s|h+)



Belief Update 1

$$b' = Z(b,a,o) \qquad replace with integral b'(s) & Z(o)a,s') \leq T(s'(s,a) b(s))$$

$$R'(b,a) = E[R(s,a)]$$

$$Particle Filter \qquad "Rejection / Unweighted" PF$$

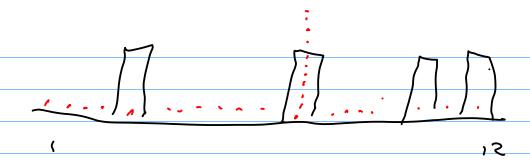
$$b' = [s], se, s, ... sk]$$

$$b'(s) = \sum_{k=1}^{n} (s,a)$$

$$|b'| = \sum_{k=1}^{n$$

output b

e replenish vandomly or with prior b besed on 0



Particle Depletion

Preventing Particle Depletion

- Add noise

- Initialize Random particles and add to b

Bootstrap Filter

Weighter PF Importance Sampling SIR

receive o $b = \overline{1}$ $5 = \overline{1}$ 5

Requires
(6(5,a)

6'(s) ≈ \(\frac{1}{2}\hat{s}_k \wk

(O(n)