# ROBOT GROCERY SHOPPING IN PARTIALLY OBSERVABLE SETTINGS

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#### OUTLINE

- 1. Background on POMDPs
- 2. Grocery shopping as planning in a POMDP
- 3. Demo!
- 4. What worked
- 5. What failed

A partially observable Markov decision process (POMDP) is a collection of objects  $(S, A, \Omega, R, T, O)$ 

*S*: state space

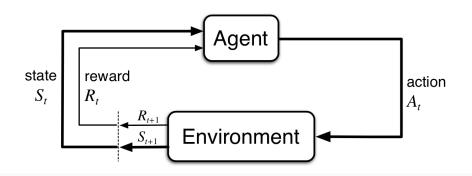
A: action space

 $\Omega$ : observation space

 $R: S \times A \rightarrow \mathbb{R}$  reward function

T: transition operator.  $T(s' \mid s, a)$  is probability of next state s' given state s and action a

*O*: observable operator.  $O(o \mid s)$  is probability of observing o given at state s



Belief-state MDP

#### Implemented MDP solvers:

Q-learning

**SARSA** 

R-MAX

Thompson sampling

There are a lot!

Function approximations with adaptive basis functions

**BOSS** 

Spectral methods

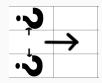
Skill chaining

. . .

#### **GROCERY SHOPPING**

#### Grid World POMDP

Uncertain movement



Can only see around current cell (partially observable)



World is not fully known beforehand

Model of how items in the same aisle correlate.

Unknown arrangement of aisles

Unknown arrangement of items within aisles

#### GROCERY SHOPPING

## *pygame* running the visuals

#### Every second:

- Agent provides next action based on current belief state
- Simulator executes action (errors may happen)
- Belief state is updated based on transition probabilities
- Belief state is updated based on observation
- Belief about the world is updated based on belief state, and observation

## Challenges:

- Markov assumption is not completely accurate
- Bias towards increasing probability of most likely states



#### OUR WORKING SOLVER

Max Probability Value Iteration:

Choose the most likely state from belief state to run value iteration

#### OUR WORKING SOLVER

#### Value iteration:

$$v_{k+1}(s) = \max_{a} \mathbb{E}[R_{t+1} + \gamma v_k(S_{t+1}) \mid S_t = s, A_t = a]$$
  
=  $\max_{a} \sum_{s'} p(s' \mid s, a)[r(s, a, s') + \gamma v_k(s')]$ 

## FAILED TASKS

Value iteration as a belief-state MDP Thompson sampling

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# PLAY WITH IT!



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