

# CMSC 170

Introduction to Artificial Intelligence

2<sup>nd</sup> Semester AY 2014-2015

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# PREVIOUSLY...

We used **problem solving techniques** (like A\*-search) to find solutions to problems, but these methods only work when environments are **fully-observable** and **deterministic**.

# WHY?

The agent thinks up a **plan/solution**, then **executes it blindly**, without sensing the environment.

# CAN WE DO THAT IN REAL LIFE?



# WE CAN'T



We thus need to **interleave**  
**planning** and **executing** to handle  
**difficult environments** like...

# 1.

## *Stochastic environments*

If the result of an action is not solely dependent on agent actions, we must **handle all possible results** after an action is performed.

# 2.

## *Multi-agent environments*

Agents must be able to **react to the actions of other agents** that may affect the environment.



# 3.

## *Partial Observability*

We might not be aware of the situation in a further section of the plan; some actions might need to be modified if certain conditions are met.

# EXAMPLE

Given a plan to go campus via Raymundo Gate:

F. O. Santos → Ruby St. → Raymundo Gate →  
CEM

The success of the solution will be **dependent on whether Raymundo Gate is open**, something that we **can't observe** when we start at **F. O. Santos**.

There may also be other factors like...

# 1.

## *Lack of/Faulty Knowledge*

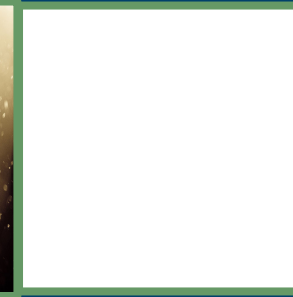
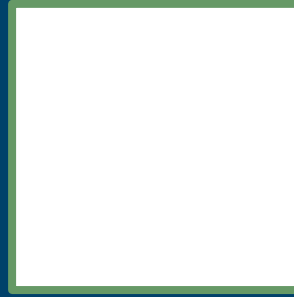
The agent may be equipped with faulty sensors, or may have been given an incomplete map.

# 2.

## *Hierarchical Plans*

Plans may **specify what** an agent should do, but **not how** to do it.

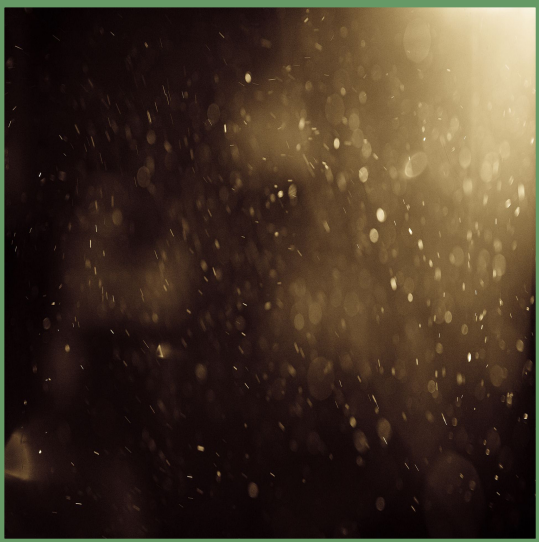
# REMEMBER...



# EXAMPLE



The robot can be in one of two positions, L or R.





# EXAMPLE



Each position  
can either have  
dust or not.

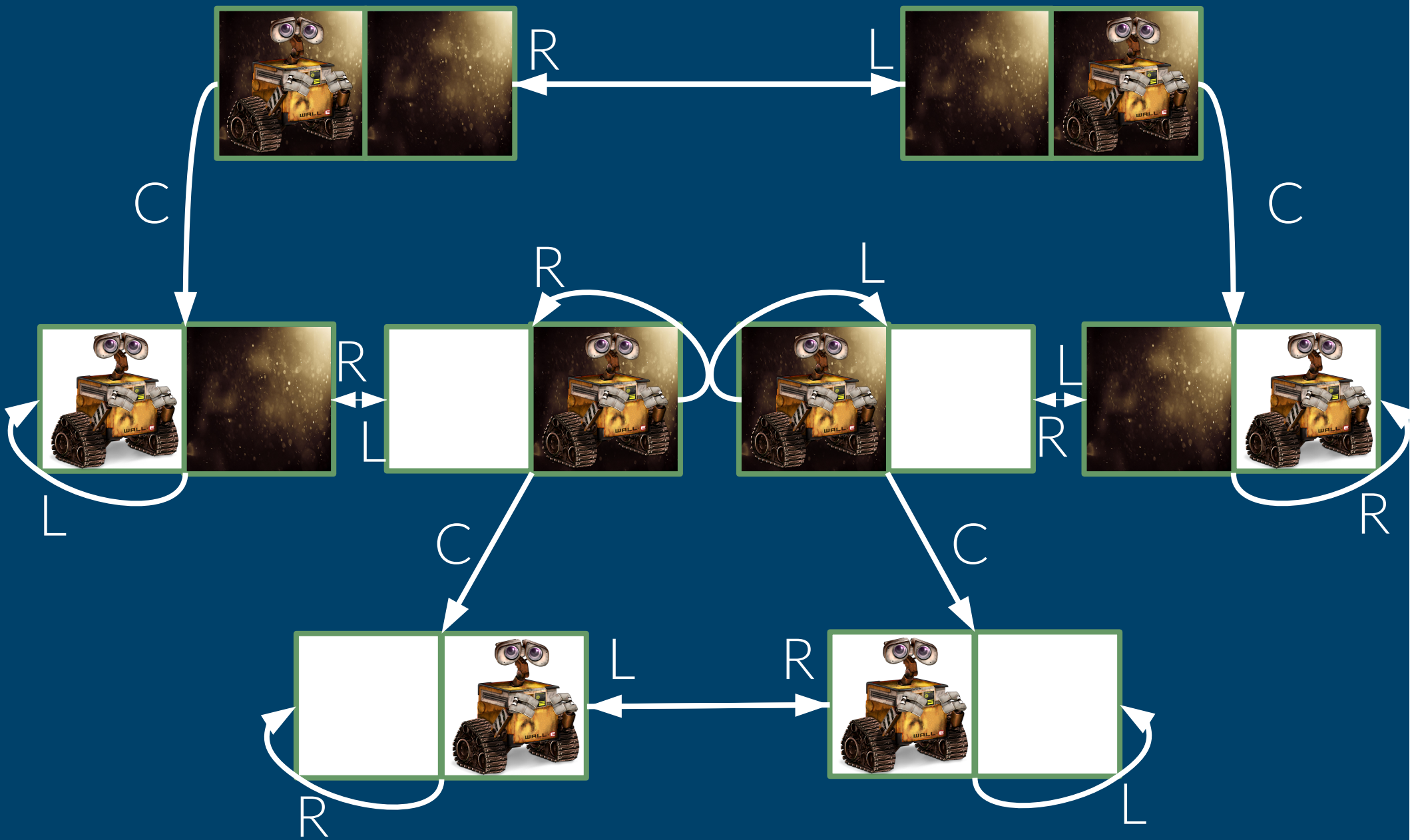


# EXAMPLE



The robot can  
clean the  
position it is in  
to remove the  
dust.





We are looking at **world states**, where each state describes exactly one, atomic state.

# BUT WHAT IF...

The sensor breaks down?

The agent can't sense its **location**.

The agent can't sense if the **location is  
dirty or not.**

Instead of looking at actual world  
states, we look at  
*belief states.*



# *Belief States*

May contain one or more world states; the agent **believes** that it is in one of them.

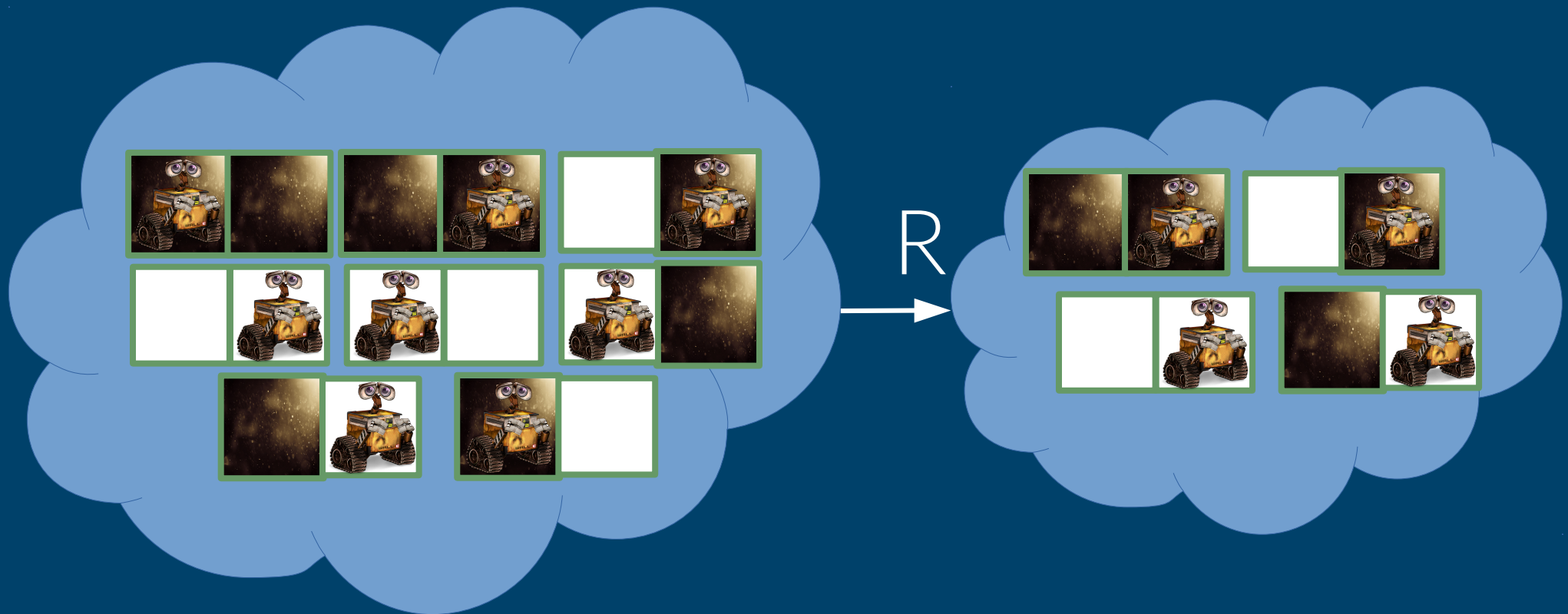


In the beginning, **we know nothing**,  
except that we are **in one of the eight**  
**states.**



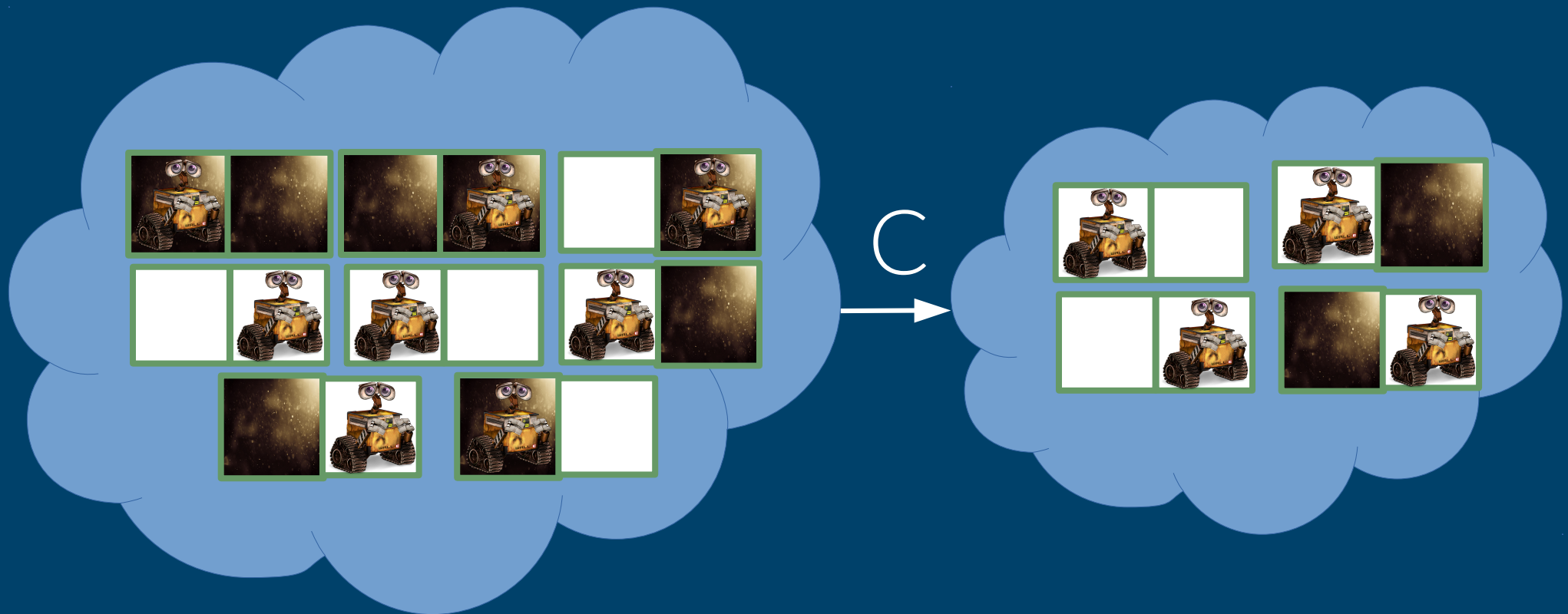


We then conduct actions to reduce the number of states in the belief state.





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# QUIZ (1/4)

Give a series of actions that will allow you to arrive at a belief state containing only states where **both rooms are clean**?

Possible actions:

R – Move to the right

L – Move to the left

C – Clean current position



# ANSWER

$L \rightarrow C \rightarrow R \rightarrow C / [L, C, R, C]$

or

$R \rightarrow C \rightarrow L \rightarrow C / [R, C, L, C]$

As much as possible, **actions** should **result** in a **belief state** with **equal** or **less** actual **world states**.

If an agent is capable of sensing local location and cleanliness, it can also **observe**, which should split the belief state into possibly several pieces.



Observation:  
Position B,  
Dirty



Observation:  
Position B,  
Clean



What if the **environment** is  
**stochastic**?

# EXAMPLE

The robot has a **good local sensor** and **vacuum**, but its wheels are slippery which **sometimes causes its attempts to move left or right to fail.**

Action:  
Move  
left

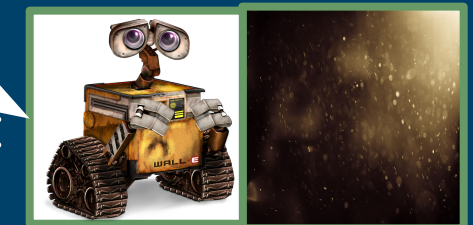
Observation:  
Position A,  
Dirty



Observation:  
Position B,  
Dirty

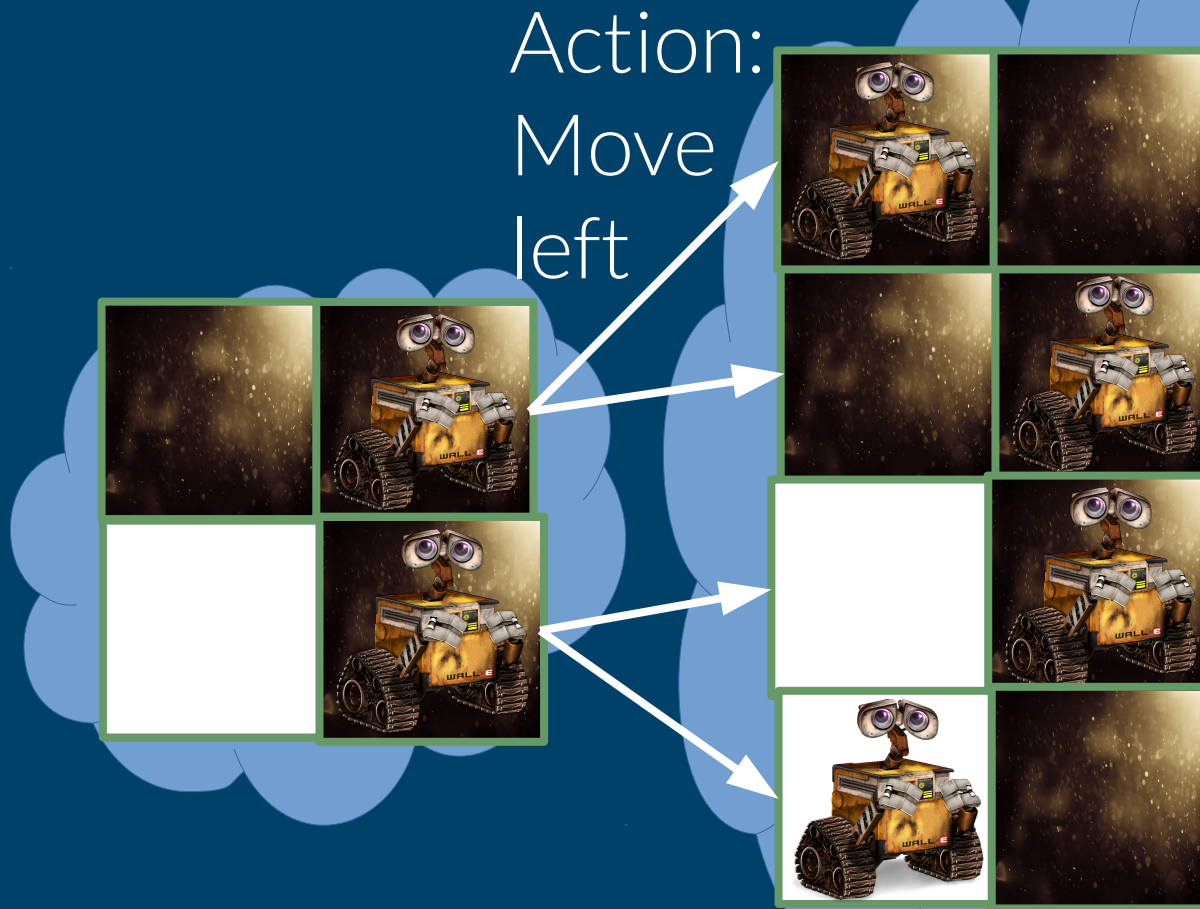


Observation:  
Position A,  
Clean

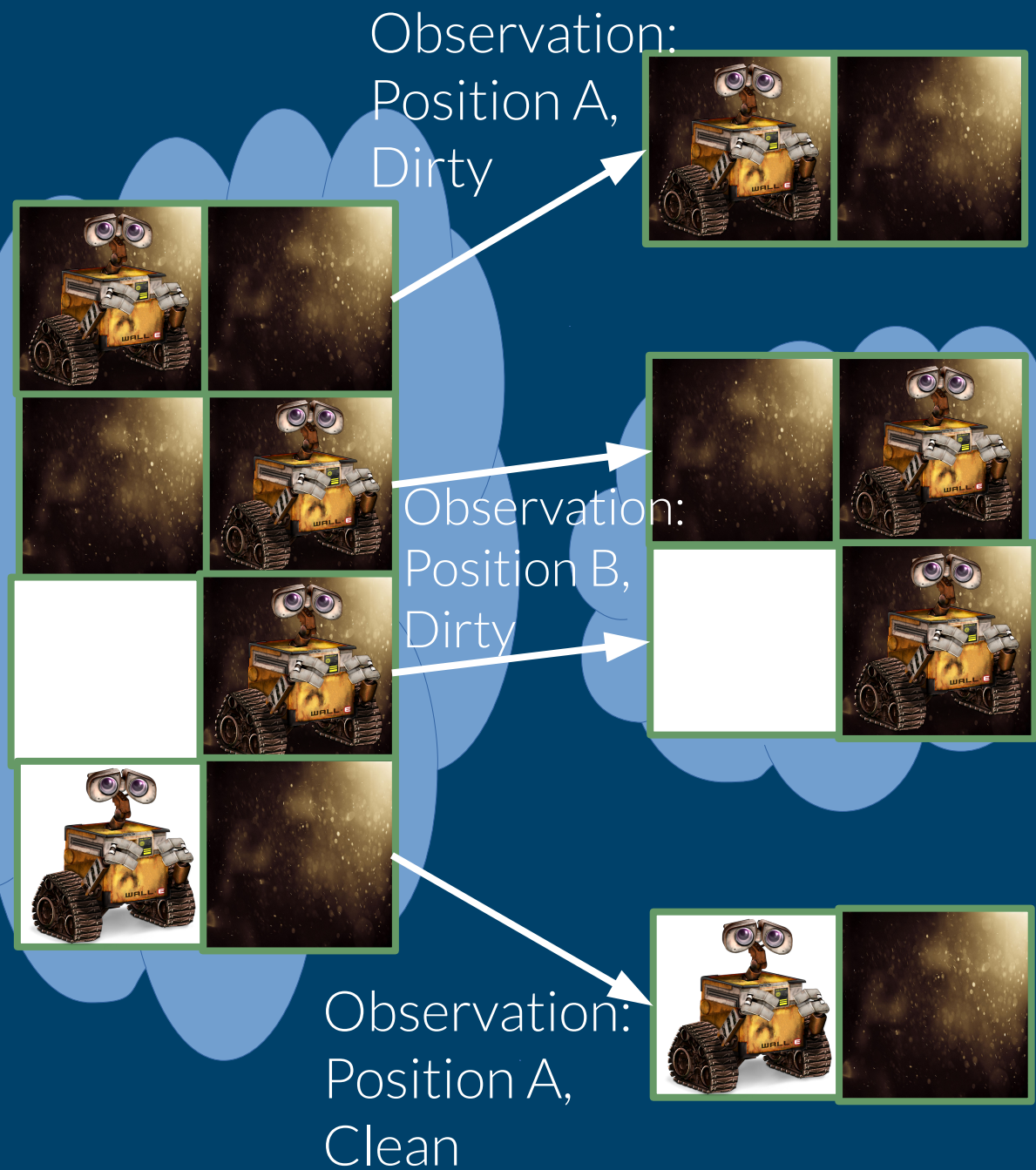




In these cases, **actions increase the size of the belief state** because the agent does not know if its action is successful or not.



Observations  
then partition  
these  
increased  
belief states  
into possibly  
several belief  
states to check  
if its action  
was successful.



In stochastic environments,  
actions increase uncertainty, while  
observations bring uncertainty  
back down.

# QUIZ (1/4)

Action:  
Move  
left

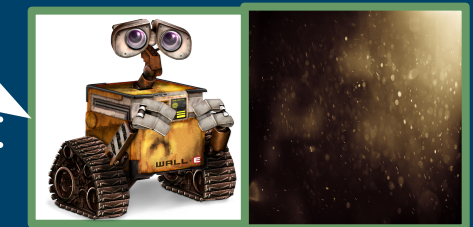
Observation:  
Position A,  
Dirty



Observation:  
Position B,  
Dirty



Observation:  
Position A,  
Clean



# QUIZ (1/4)

Will each of the following plans always result in two clean rooms (the agent has a good local sensor and vacuum, but slippery wheels)? Answer yes or no only.

1.[C, R, C]

2.[R, C, L, C]

3.[C, R, R, C]

4.[C, R, C, R, C]

Possible actions:

C – Clean

R – Move right

L – Move left

# ANSWERS

1.No

2.No

3.No

4.No

Given a **stochastic environment**, plans with **finite steps** are **not guaranteed** to find the goal belief/world state; we must allow notation for **infinite steps**.

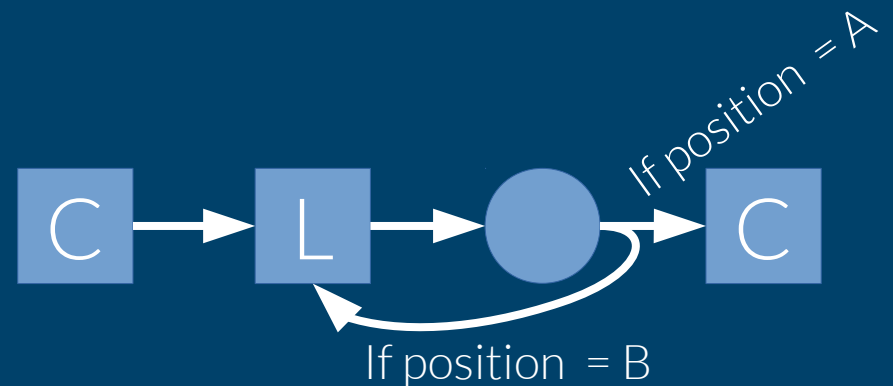


# EXAMPLE



Finite Sequence:  
[C, R, C]

Infinite Sequence:



[S, while position B: L, C]



In stochastic environments, there is no way to infer the finite number of steps required to eventually reach the goal.