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# Hybrid Declarative-Imperative Representations for Hybrid Discrete-Continuous Decision-Making



Jiayuan Mao



Joshua B. Tenenbaum



Tomás Lozano-Pérez



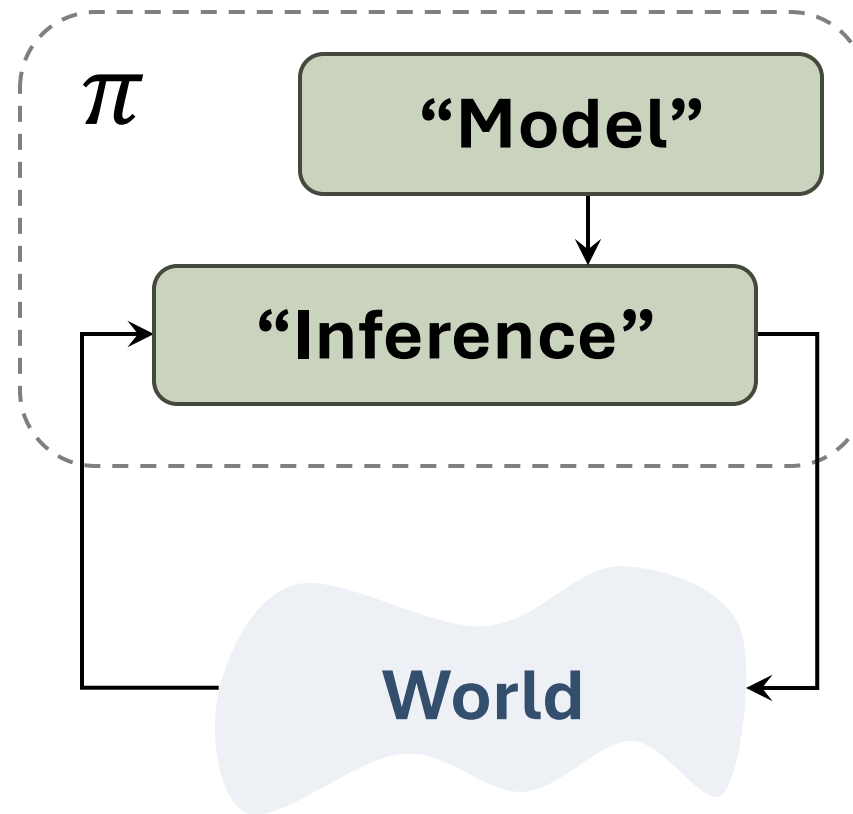
Leslie Pack Kaelbling



Massachusetts  
Institute of  
Technology

jiayuanm.com  
jiayuanm@mit.edu

# Structures of the “Robot Brain”



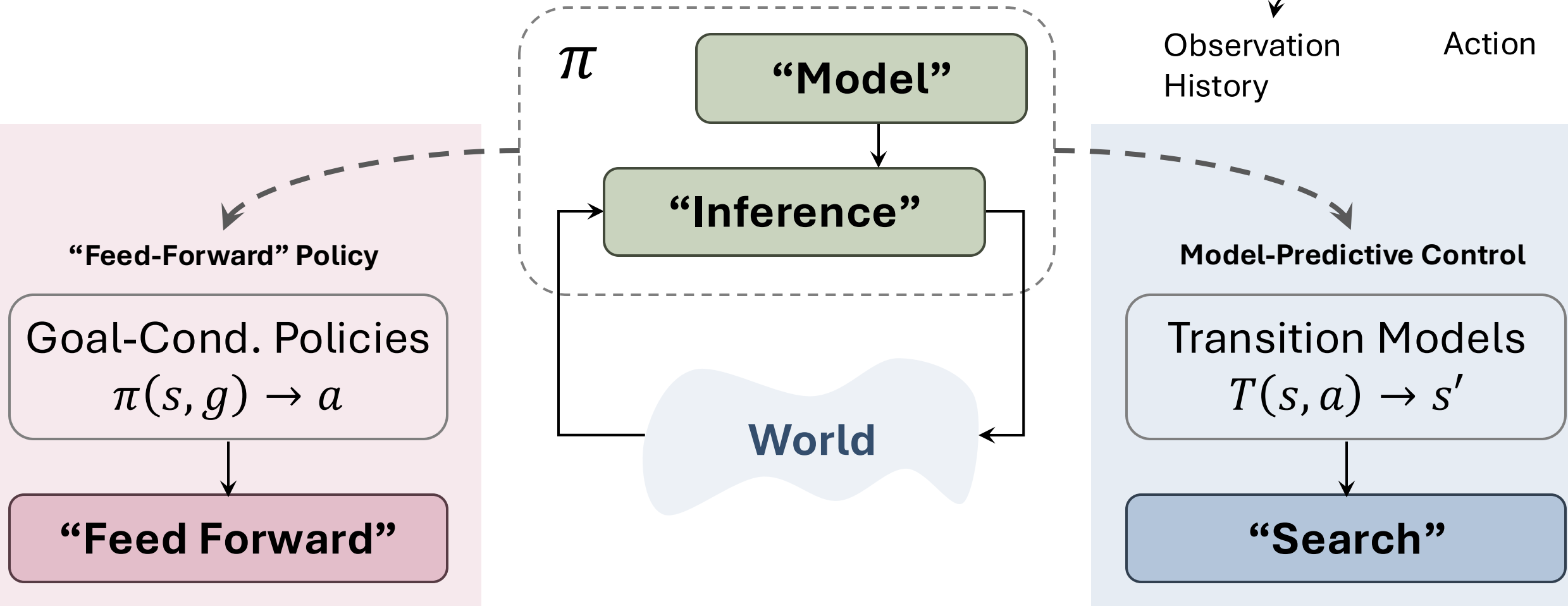
$$\pi: \underbrace{(o, a)^*}_{\substack{\text{Observation} \\ \text{History}}} \rightarrow a \downarrow \text{Action}$$

# Structures of the “Robot Brain”

$$\pi: \underbrace{(o, a)^*}_{\substack{\text{Observation} \\ \text{History}}} \rightarrow a$$

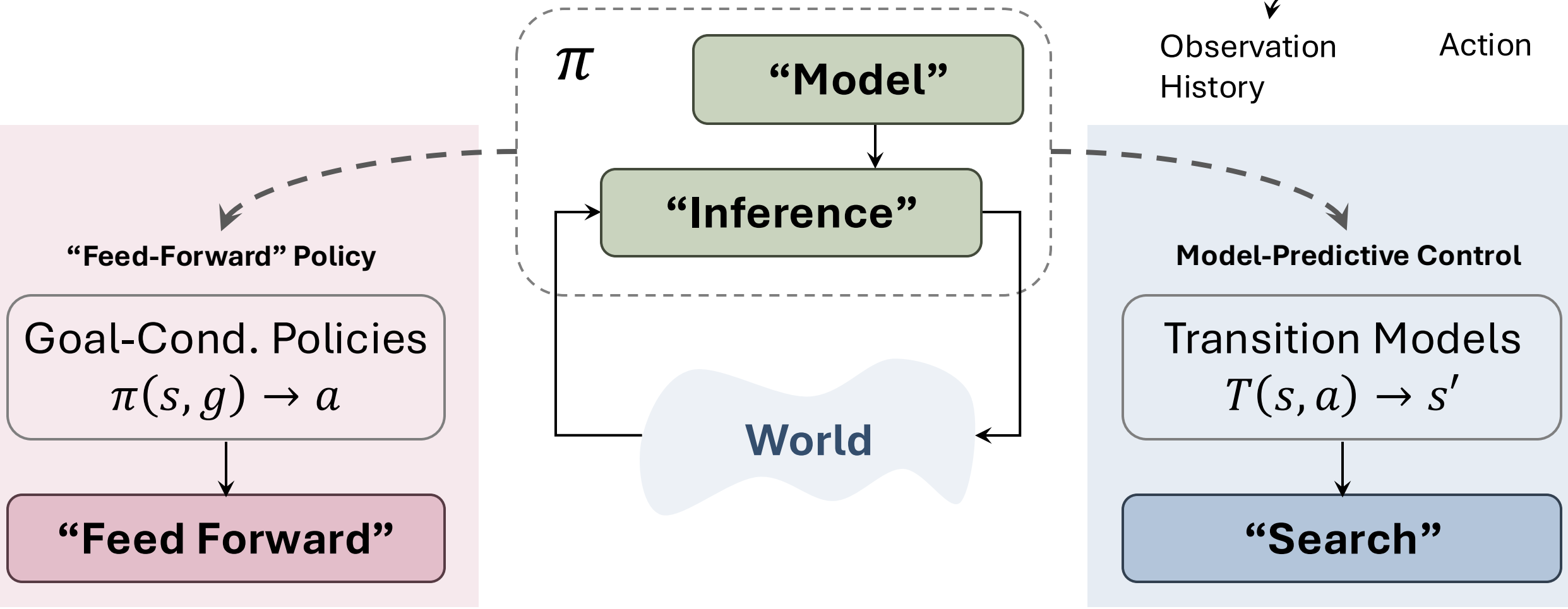
Observation  
History

Action



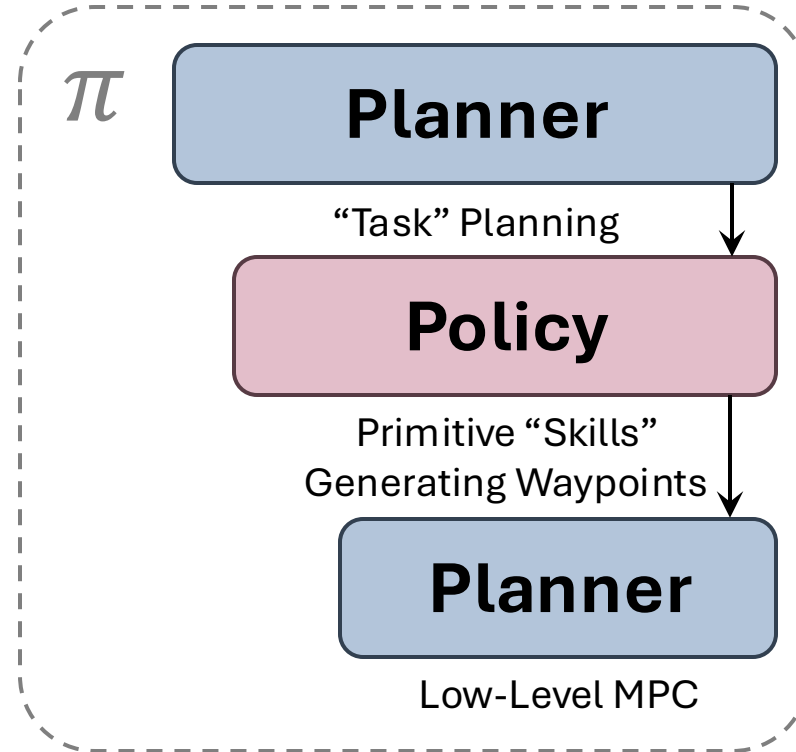
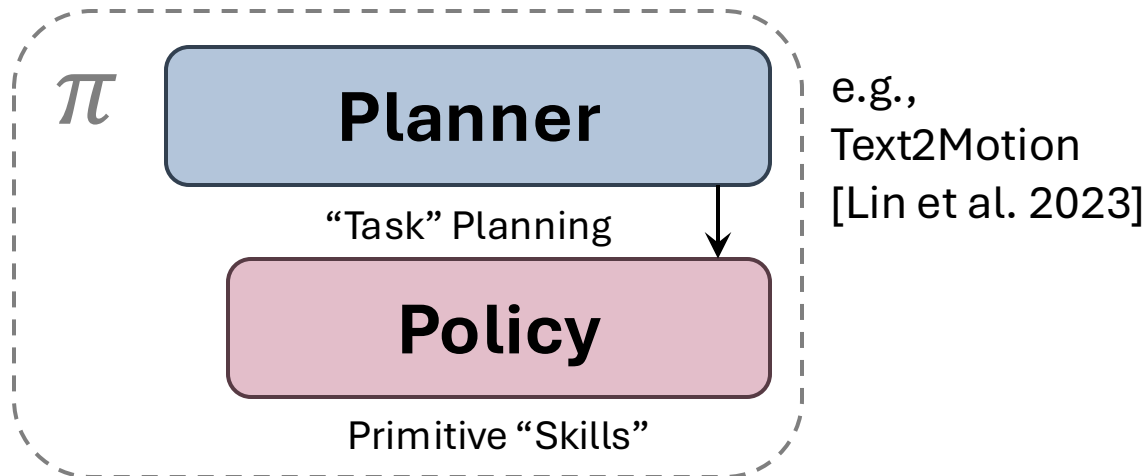
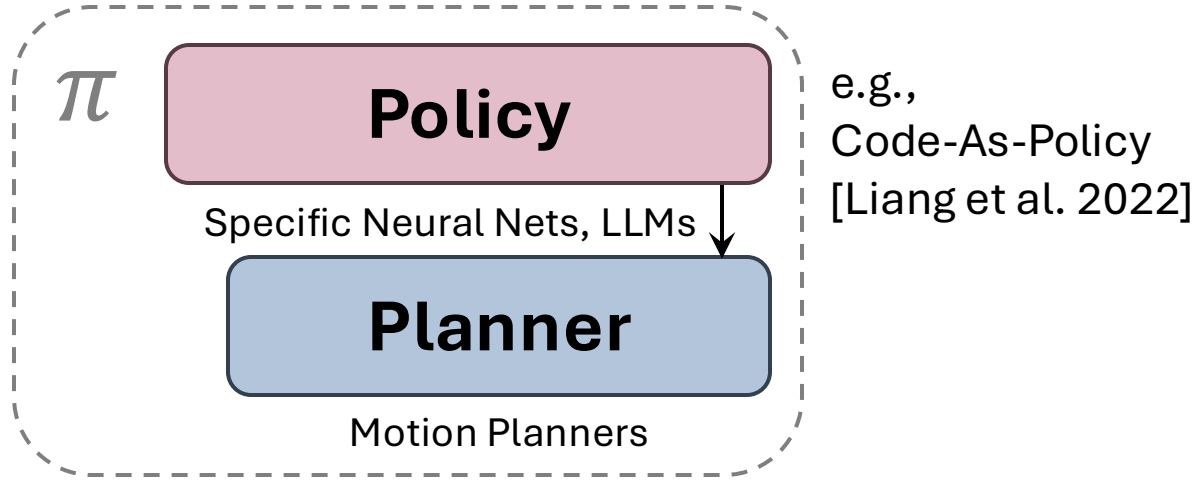
# Observations: “We Need Both”

$$\pi: \underbrace{(o, a)^*}_{\substack{\text{Observation} \\ \text{History}}} \rightarrow a \downarrow \text{Action}$$



The choice between policy and model depends on the context and the task  
Many times, they need to be combined

# A Broad Class of “Hybrid” Systems



# Open Research Questions

## Theory

How to **mathematically** describe all these combinations and their trade-offs (e.g. the complexity)?

## Practice

How to flexibly **mix-and-match** all these modules to build efficient and effective systems?

# The Continuous Spectrum of Hybrid Systems

**Today's talk:** a unified theory, starting with a “programming language”

## Imperative Representations

Goal-Cond. Policies

$$\pi(s, g) \rightarrow a$$



**“Feed Forward”**

## Declarative Representations

Transition Models

$$\pi(s, a) \rightarrow s'$$



**“Search”**

# NAMO in the Crow Description Language: Basic Primitives

**State:** the state is represented as a set of objects and relational features

```
object A, B, C: object
```

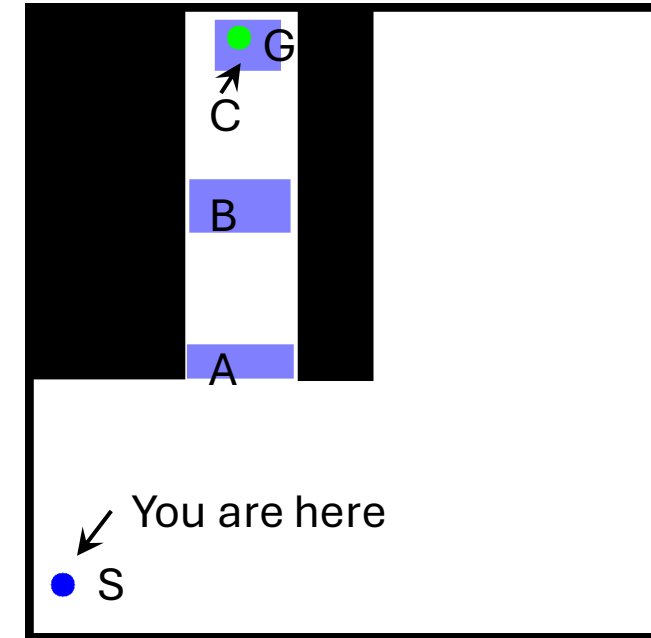
```
feature shape_of(o: object) -> vector
```

```
feature pos_of(o: object) -> vector
```

**Primitive Action:** parameterized “low-level” controllers

```
controller move_path(t: list[vector])
```

```
controller attach(o: object)
```



“Navigation Among Movable Obstacles”

Reif and Sharir, 1985

Wilfong, 1988

Stilman and Kuffner, 2005

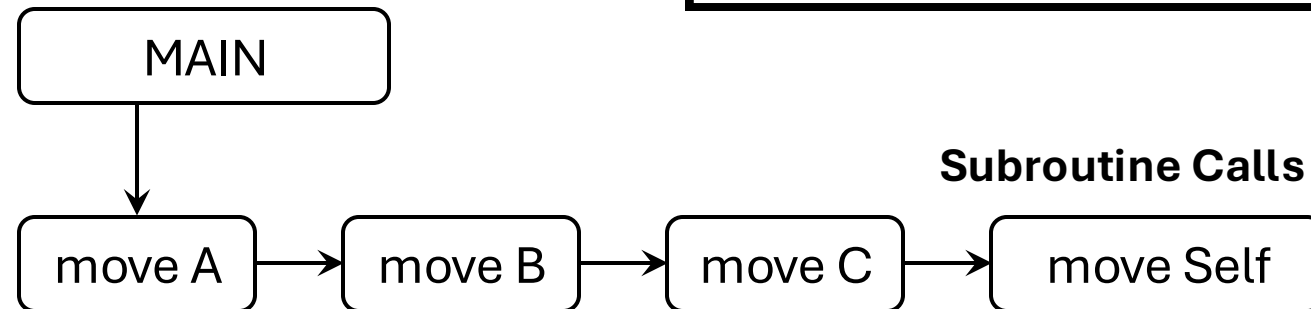
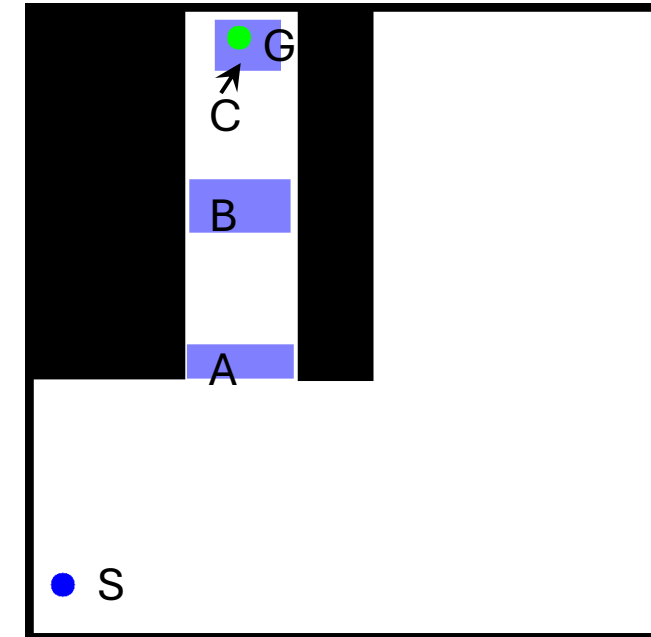


# Directly Programmed Solution

## Imperative

```
global_goal: agent_pos() == (270, 50)

behavior goto_v0(G):
  goal: agent_pos() == G
  body:
    achieve pos_of(A) == (500, 100)
    achieve pos_of(B) == (500, 300)
    achieve pos_of(C) == (500, 500)
    let path = find_path(agent_pos(), G)
    do move_path(path)
```



Like “Behavior Trees”

Mateas and Stern. 2002. “A Behavior Language for story-based believable agents”  
Bagnell et al. 2012. “An Integrated System for Autonomous Robotics Manipulation”  
Colledanchise and Ögren. 2018 “Behavior Trees in Robotics and AI”



# Adding Flexible Ordering

Imperative +Variables +Ordering

```
global_goal: agent_pos() == (270, 50)
```

```
behavior goto_v1(G: vector):
```

```
  goal: agent_pos() == G
```

```
  body:
```

```
    bind path = find_path(agent_pos(), G)
```

```
    unordered:
```

```
      achieve not_blocking(A, path)
```

```
      achieve not_blocking(B, path)
```

```
      achieve not_blocking(C, path)
```

```
    do move_path(path)
```

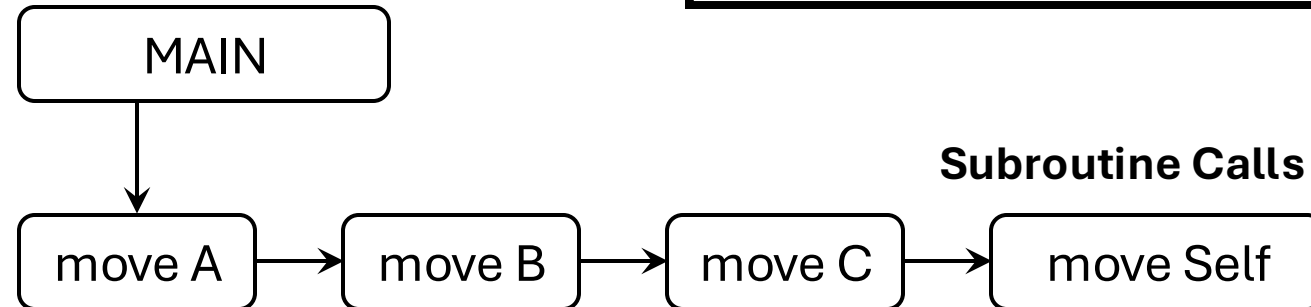
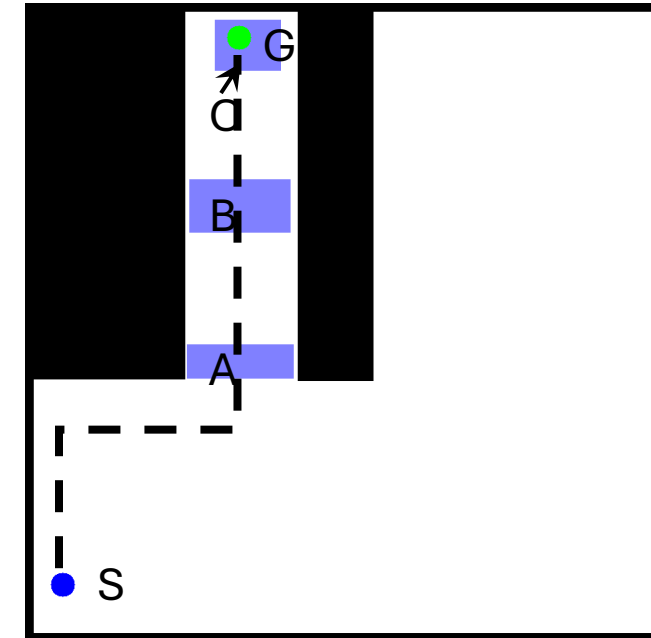
```
behavior move_away(x: object, path):
```

```
  goal: not_blocking(x, p)
```

```
  body:
```

```
    bind new_p: valid_pos(x, new_p)
```

```
    ...
```



# Adding Flexible Ordering

Imperative

+Variables

+Ordering

```
global_goal: agent_pos() == (270, 50)
```

```
behavior goto_v1(G: vector):
```

```
  goal: agent_pos() == G
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```
  body:
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    bind path = find_path(agent_pos(), G)
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    unordered:
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```
      achieve not_blocking(A, path)
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```
    do move_path(path)
```

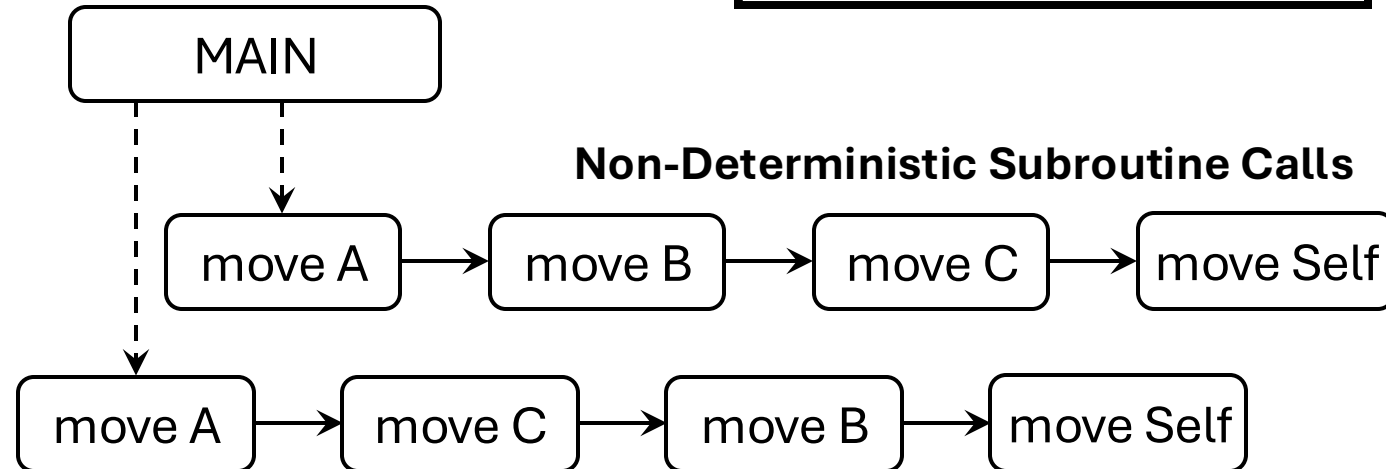
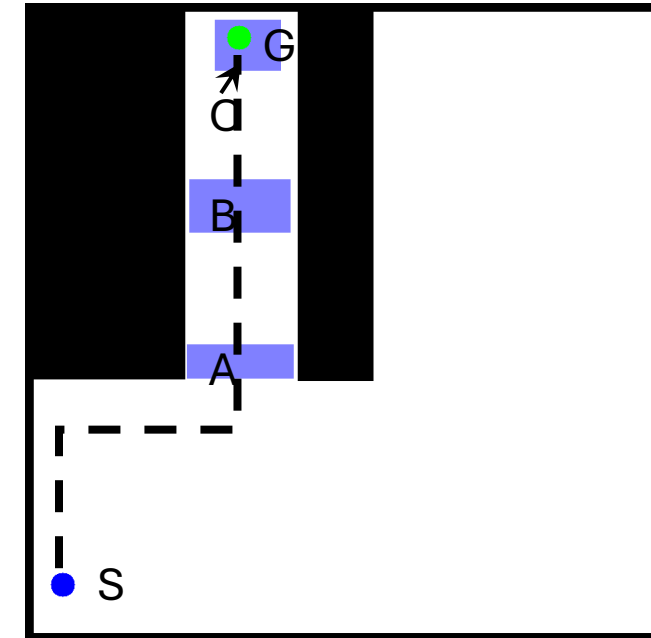
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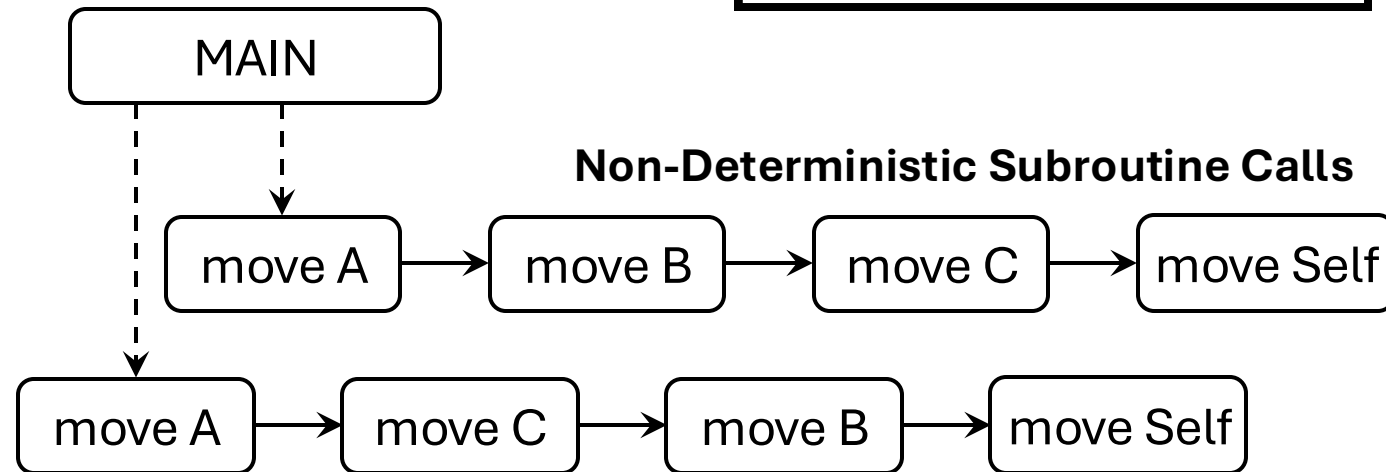
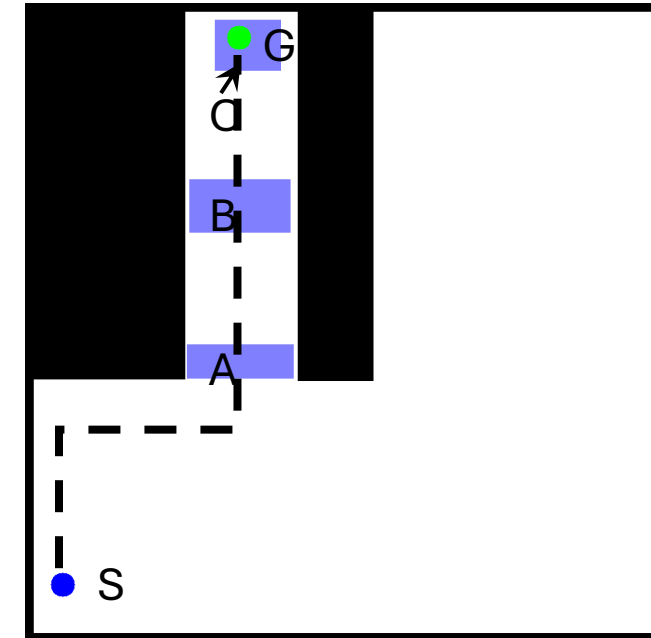
```
  body:
```

```
    assert reachable(x)
```

```
    bind new_p: valid_pos(x, new_p)
```

```
    ...
```

```
  eff: pos_of(x) = new_p
```



# Adding Flexible Ordering

Imperative

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global_goal: agent_pos() == (270, 50)
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behavior goto_v1(G: vector):
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```
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```

```
    bind path = find_path(agent_pos(), G)
```

```
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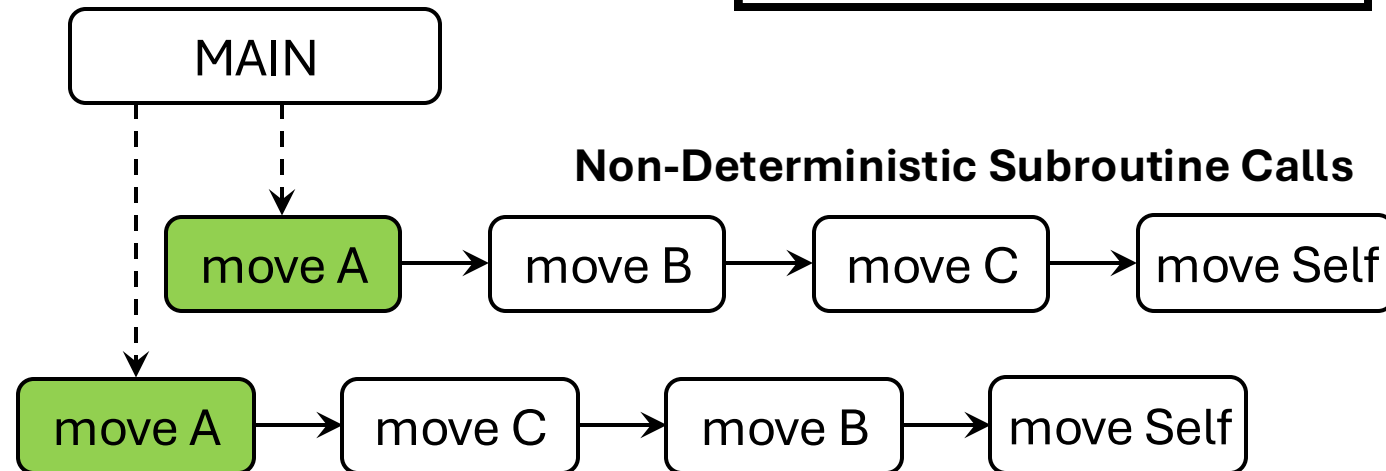
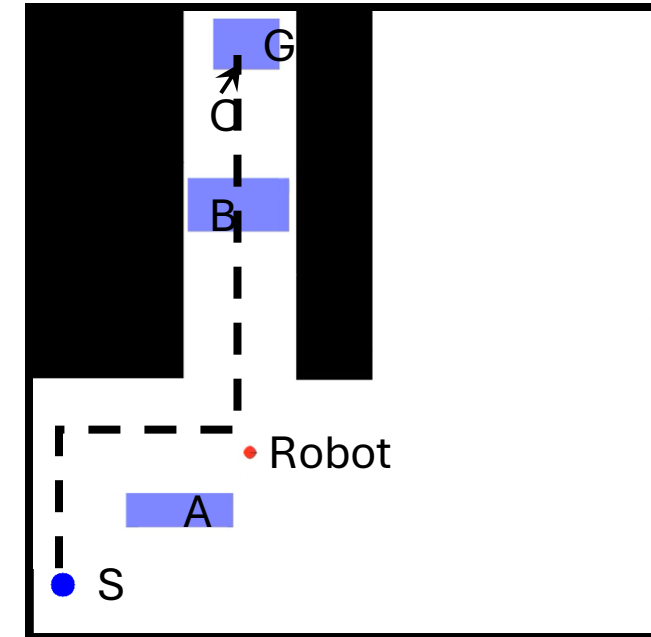
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```
    ...
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# Adding Flexible Ordering

Imperative +Variables +Ordering

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behavior goto_v1(G: vector):
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  goal: agent_pos() == G
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```
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```

```
    bind path = find_path(agent_pos(), G)
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      achieve not_blocking(C, path)
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    do move_path(path)
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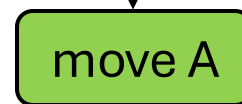
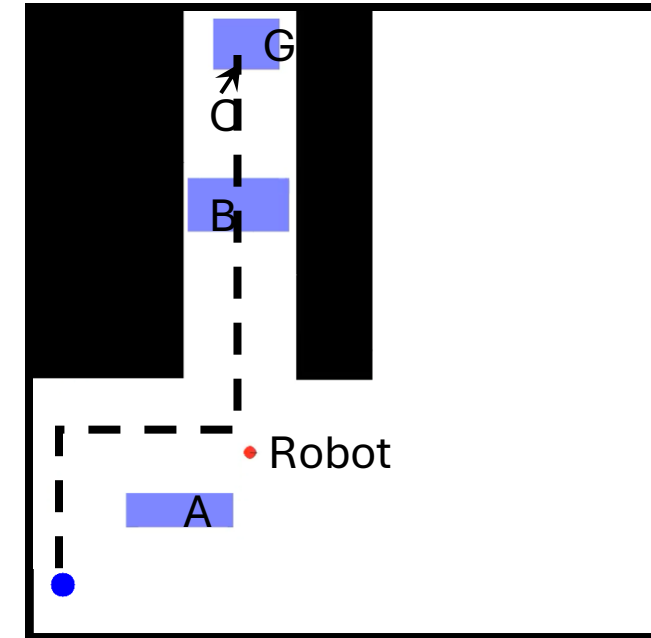
```
  body:
```

```
    assert reachable(x)
```

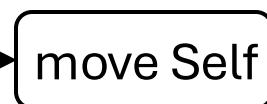
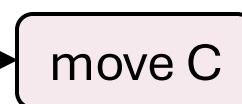
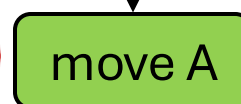
```
    bind new_p: valid_pos(x, new_p)
```

```
    ...
```

```
  eff: pos_of(x) = new_p
```



Non-Deterministic Subroutine Calls



# Adding Flexible Ordering

Imperative

+Variables

+Ordering

```
global_goal: agent_pos() == (270, 50)
```

```
behavior goto_v1(G: vector):
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```
  goal: agent_pos() == G
```

```
  body:
```

```
    bind path = find_path(agent_pos(), G)
```

```
    unordered:
```

```
      achieve not_blocking(A, path)
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      achieve not_blocking(C, path)
```

```
    do move_path(path)
```

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behavior move_away(x: object, path):
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```
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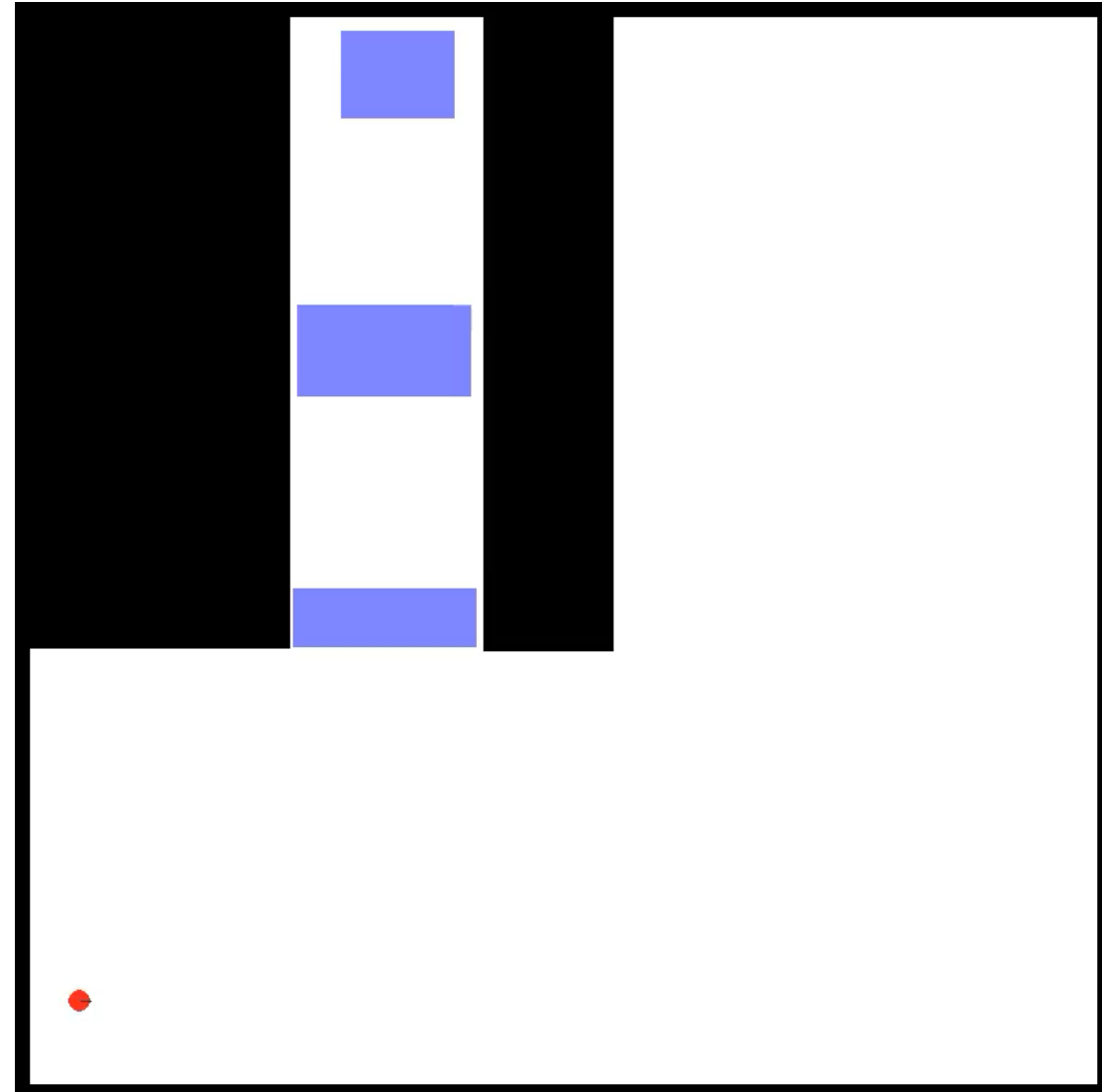
```
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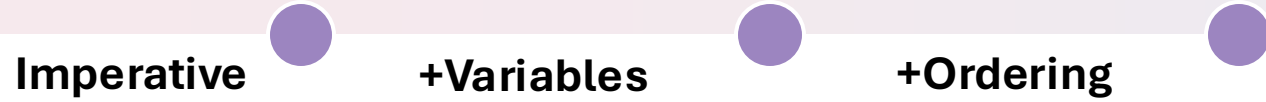
```
    ...
```

```
  eff: pos_of(x) = new_p
```





# The Spectrum Between Imperative and Declarative



**Insight 1:** Behaviors = Generators of “non-deterministic subroutine calls”  
+ Verifiers based on causal models

# Adding Flexible Ordering

Imperative +Variables +Ordering

```
global_goal: agent_pos() == (270, 50)
```

```
behavior goto_v1(G: vector):
```

```
  goal: agent_pos() == G
```

```
  body:
```

```
    bind path = find_path(agent_pos(), G)
```

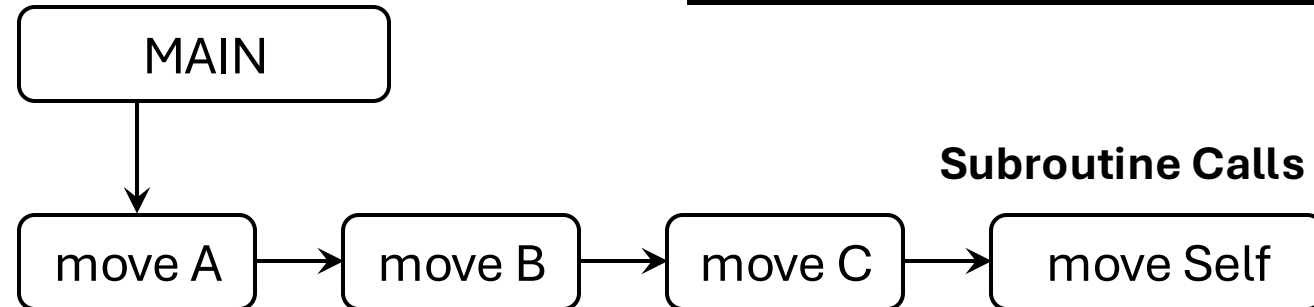
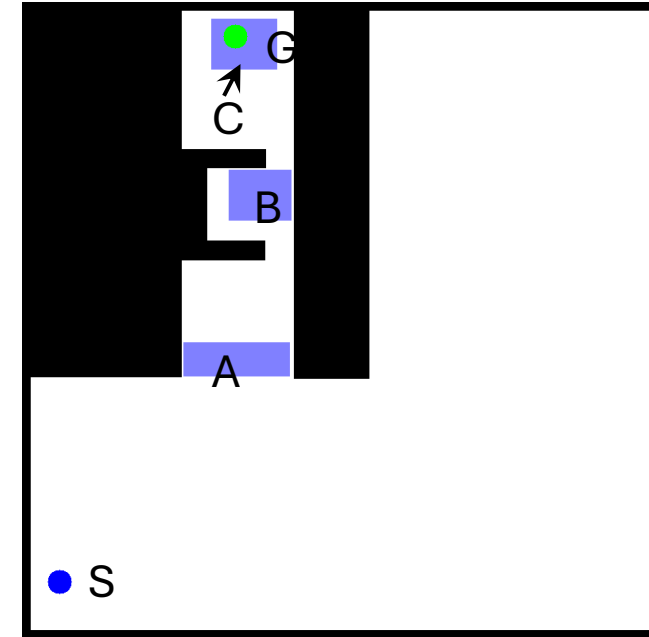
```
    unordered:
```

```
      achieve not_blocking(A, path)
```

```
      achieve not_blocking(B, path)
```

```
      achieve not_blocking(C, path)
```

```
    do move_path(path)
```



# Adding More Recursive Subroutines

Imperative

+Variables

+Ordering

```
global_goal: agent_pos() == (270, 50)
```

```
behavior goto_v2(G: vector):
```

```
  goal: agent_pos() == G
```

```
  body:
```

```
    bind waypoint: vector
```

```
    achieve agent_pos() == waypoint
```

```
    bind path = find_path(agent_pos(), G)
```

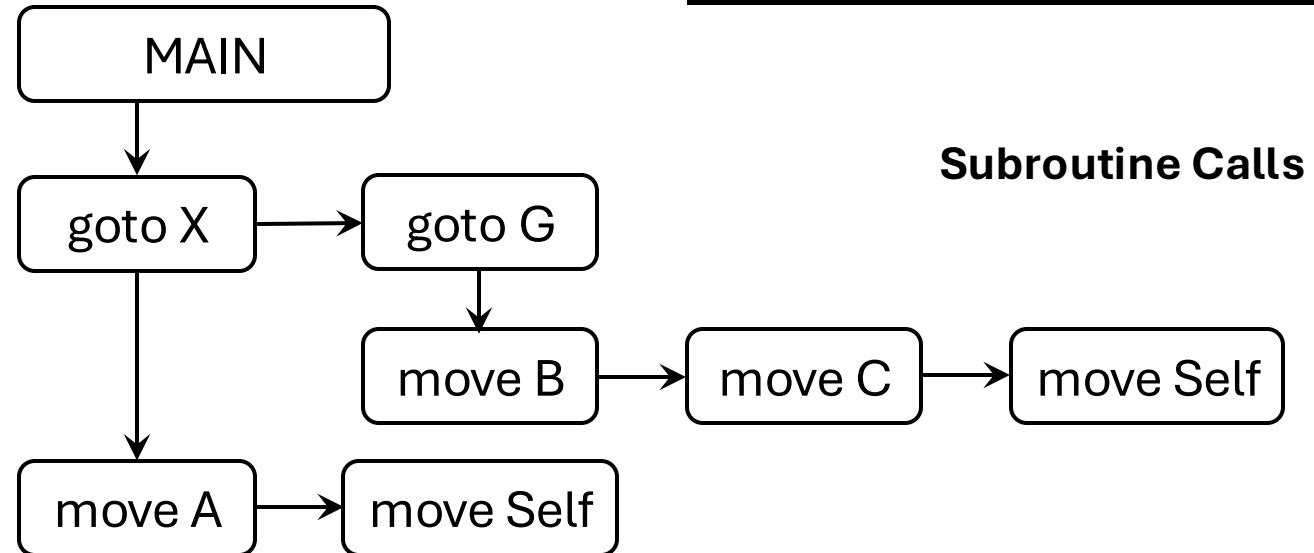
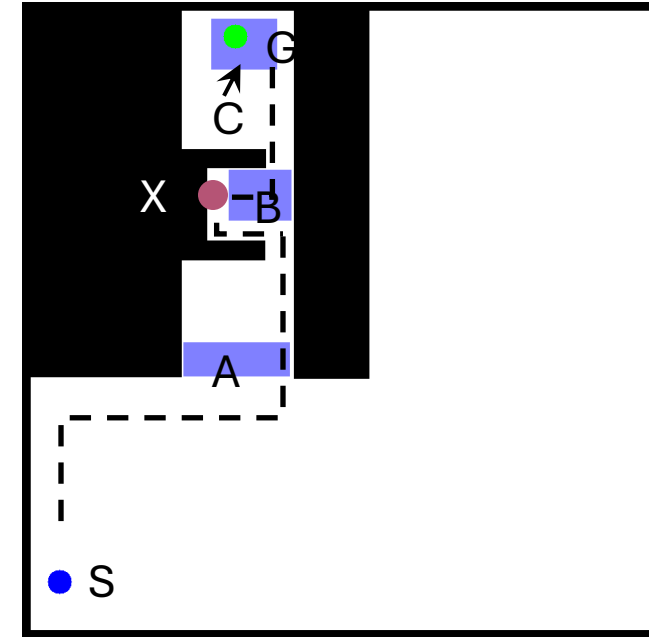
```
    unordered:
```

```
      achieve not_blocking(A, path)
```

```
      achieve not_blocking(B, path)
```

```
      achieve not_blocking(C, path)
```

```
    do move_path(path)
```



# Adding More Recursive Subroutines

Imperative

+Variables

+Ordering

```
global_goal: agent_pos() == (270, 50)
```

```
behavior goto_v2(G: vector):
```

```
  goal: agent_pos() == G
```

```
  body:
```

```
    [bind waypoint: vector
```

```
    achieve agent_pos() == waypoint
```

```
    bind path = find_path(agent_pos(), G)
```

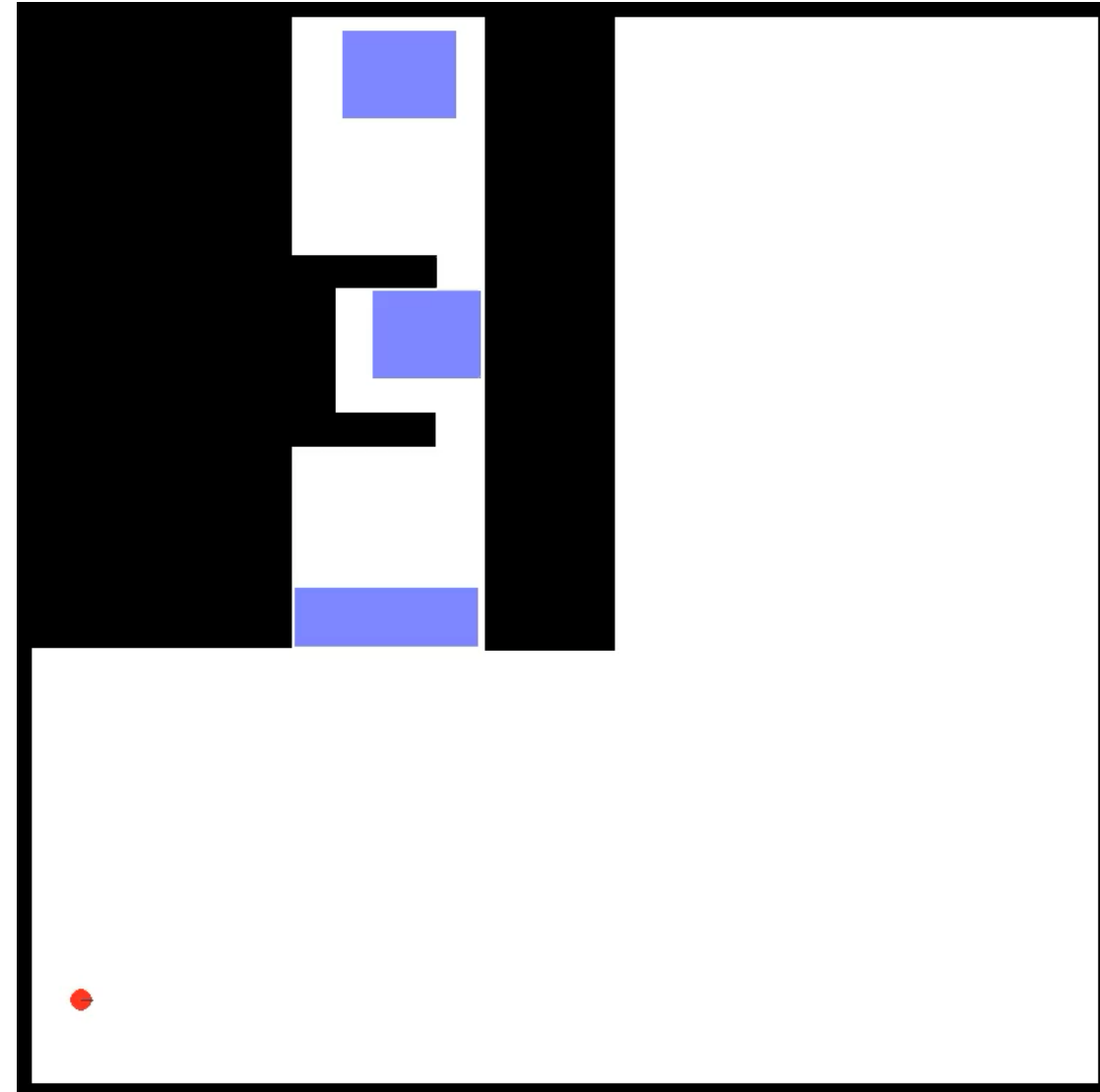
```
    unordered:
```

```
      achieve not_blocking(A, path)
```

```
      achieve not_blocking(B, path)
```

```
      achieve not_blocking(C, path)
```

```
  do move_path(path)
```



# Adding Flexible Promotion

Imperative

+Variables

+Ordering

+Promotion

```
global_goal: agent_pos() == (270, 50)
```

```
behavior goto_v2(G: vector):
```

```
  goal: agent_pos() == G
```

```
  body:
```

```
    bind waypoint: vector
```

```
    achieve agent_pos() == waypoint
```

```
    bind path = find_path(agent_pos(), G)
```

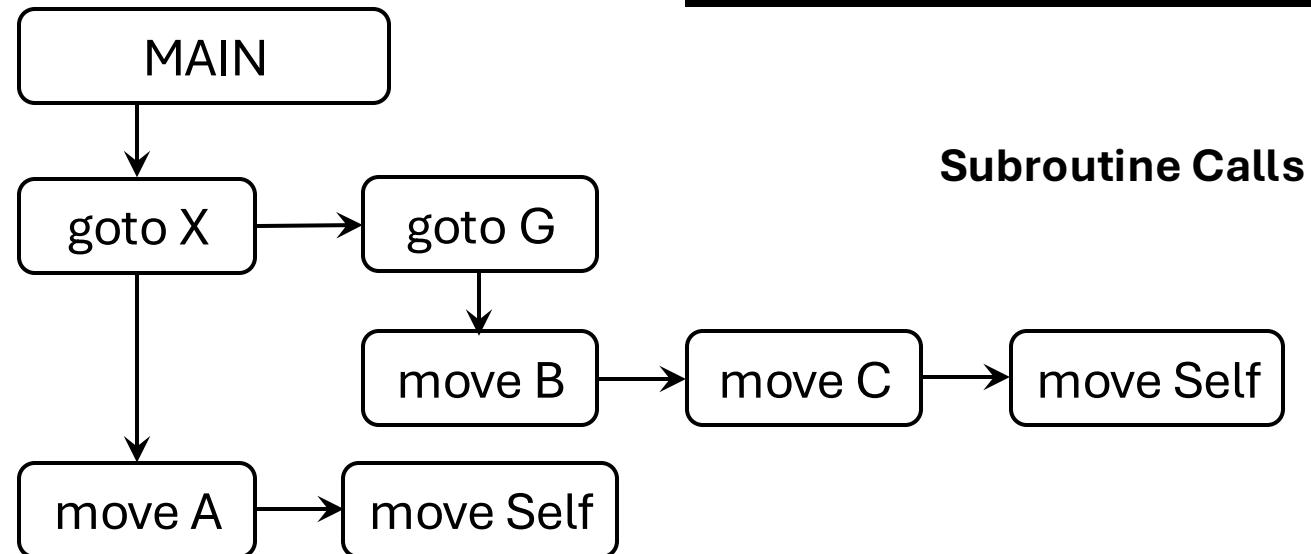
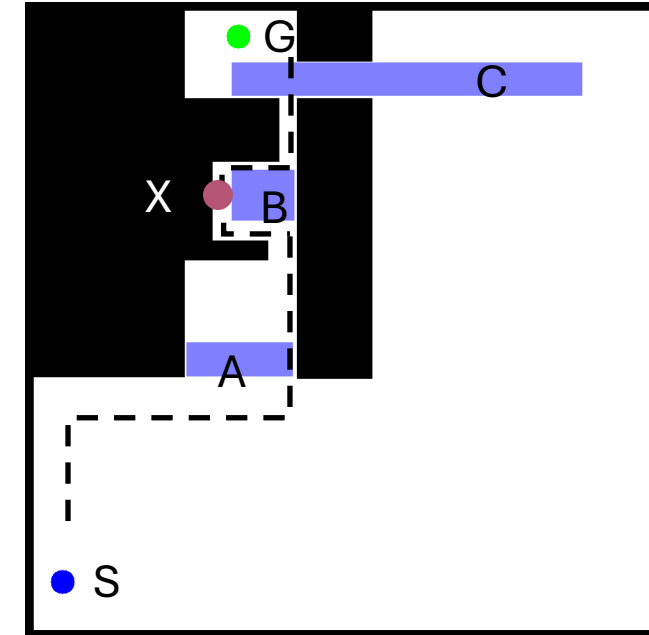
```
    unordered:
```

```
      achieve not_blocking(A, path)
```

```
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```

```
      achieve not_blocking(C, path)
```

```
    do move_path(path)
```





# Adding Flexible Promotion

Imperative

+Variables

+Ordering

+Promotion

```
global_goal: agent_pos() == (270, 50)
```

```
behavior goto_v3(G: vector):
```

```
  goal: agent_pos() == G
```

```
  body:
```

```
    bind waypoint: vector
```

```
    achieve agent_pos() == waypoint
```

```
    bind path = find_path(agent_pos(), G)
```

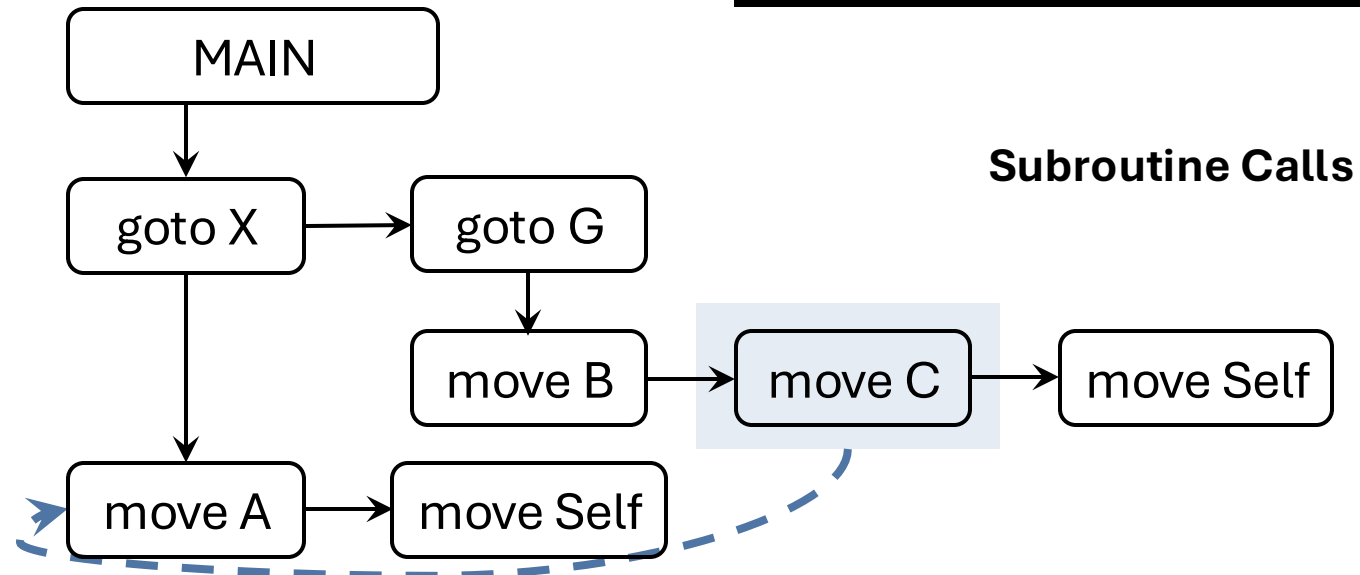
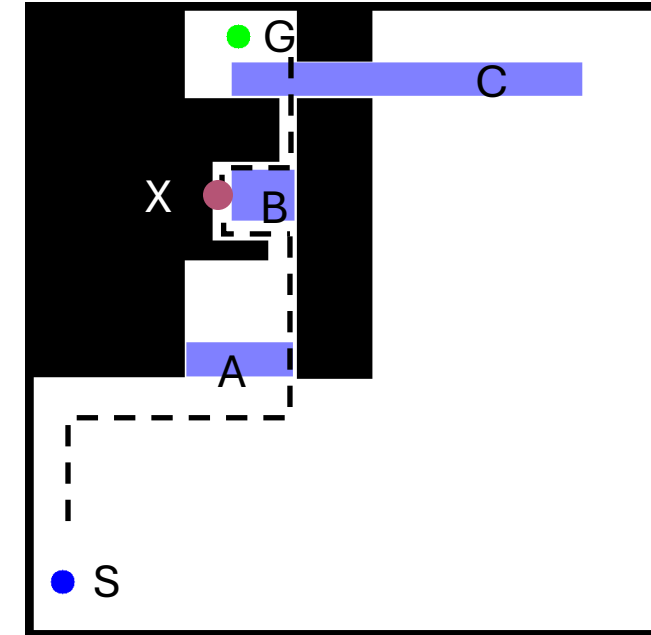
```
    promotable unordered:
```

```
      achieve not_blocking(A, path)
```

```
      achieve not_blocking(B, path)
```

```
      achieve not_blocking(C, path)
```

```
    do move_path(path)
```



# Adding Flexible Promotion

Imperative

+Variables

+Ordering

+Promotion

MAIN

goto X

goto G

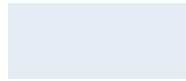
move A

move Self

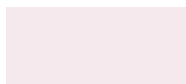
move B

move C

move Self



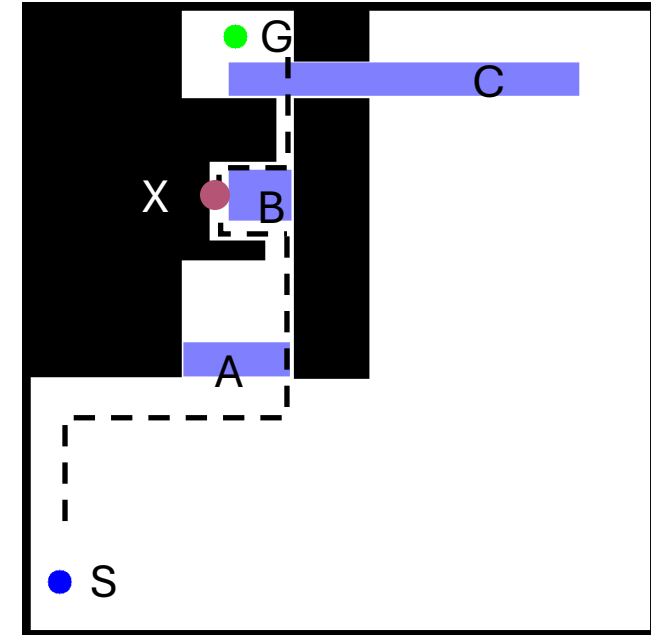
Promotable



Serialized

Ordered

Ordered





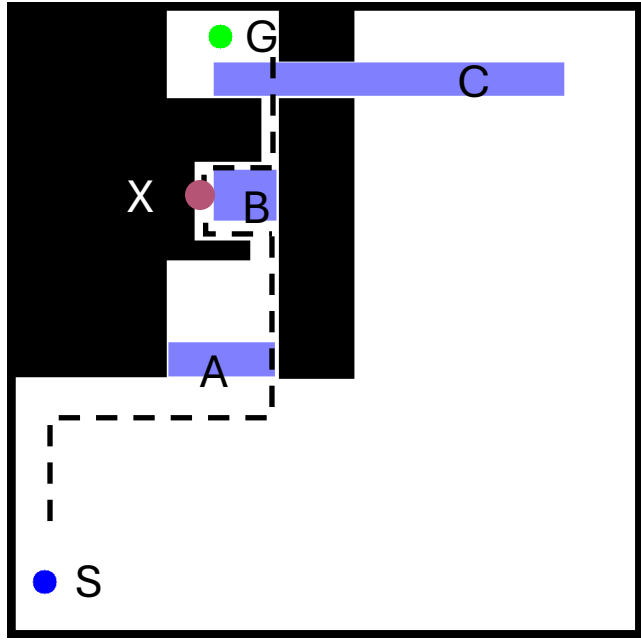
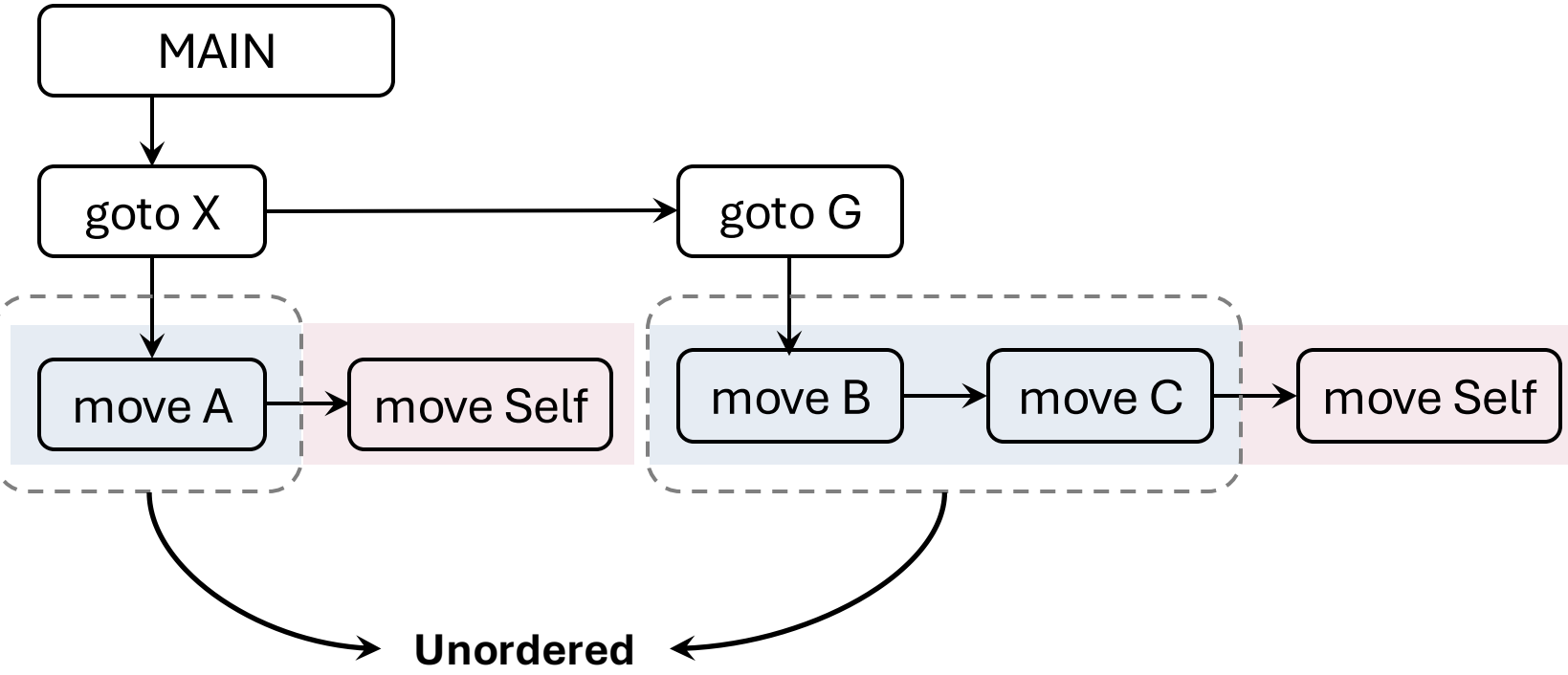
# Adding Flexible Promotion

Imperative

+Variables

+Ordering

+Promotion



# Adding Flexible Promotion

Imperative

+Variables

+Ordering

+Promotion

MAIN

goto X

goto G

move A

move Self

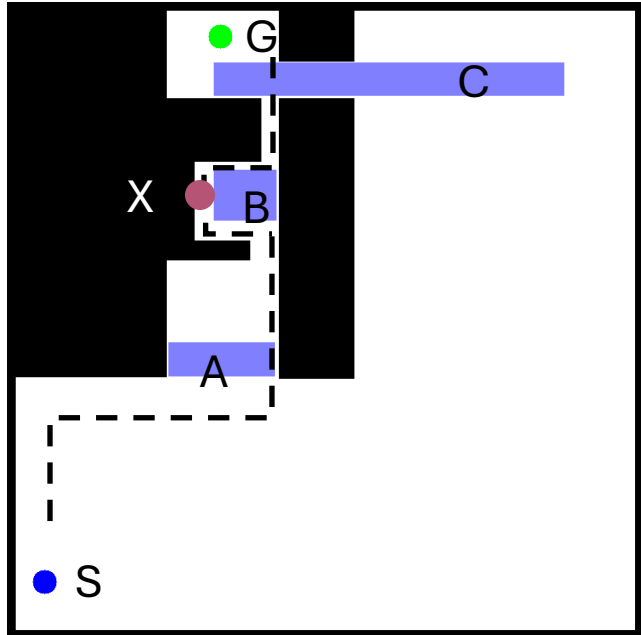
move B

move C

move Self

move to X

move to G



# Adding Flexible Promotion

Imperative

+Variables

+Ordering

+Promotion

MAIN

goto X

goto G

move A

move Self

move B

move C

move Self

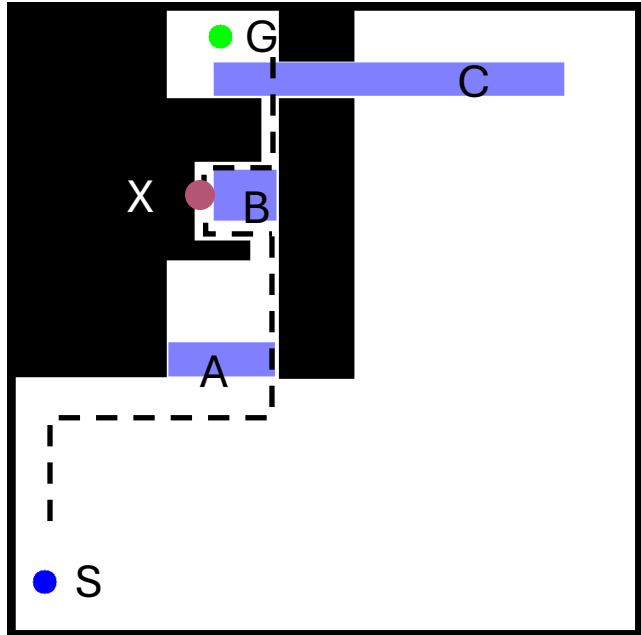
move A

move to X

move B

move C

move to G



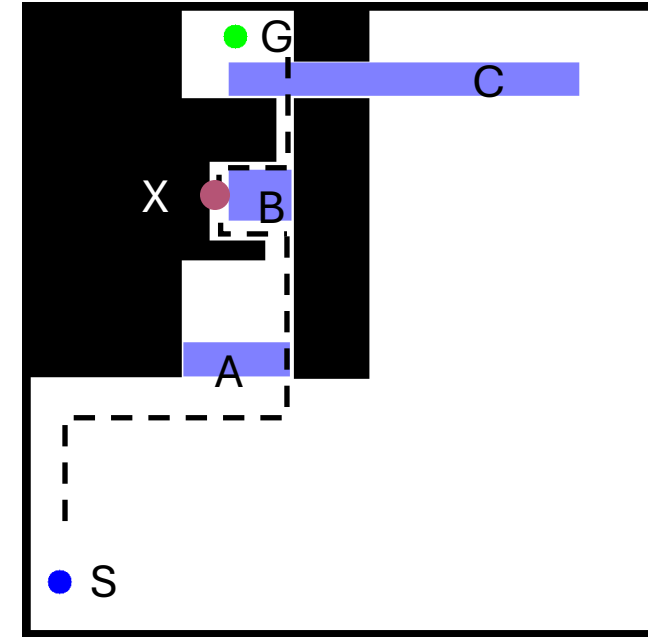
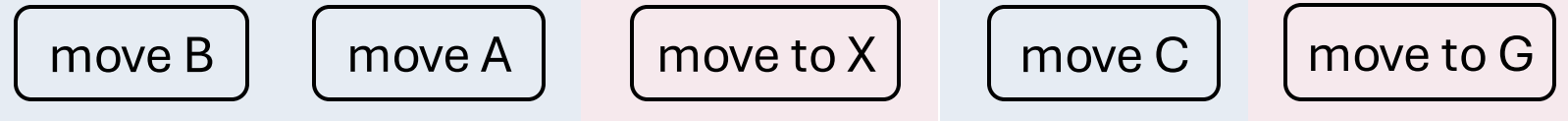
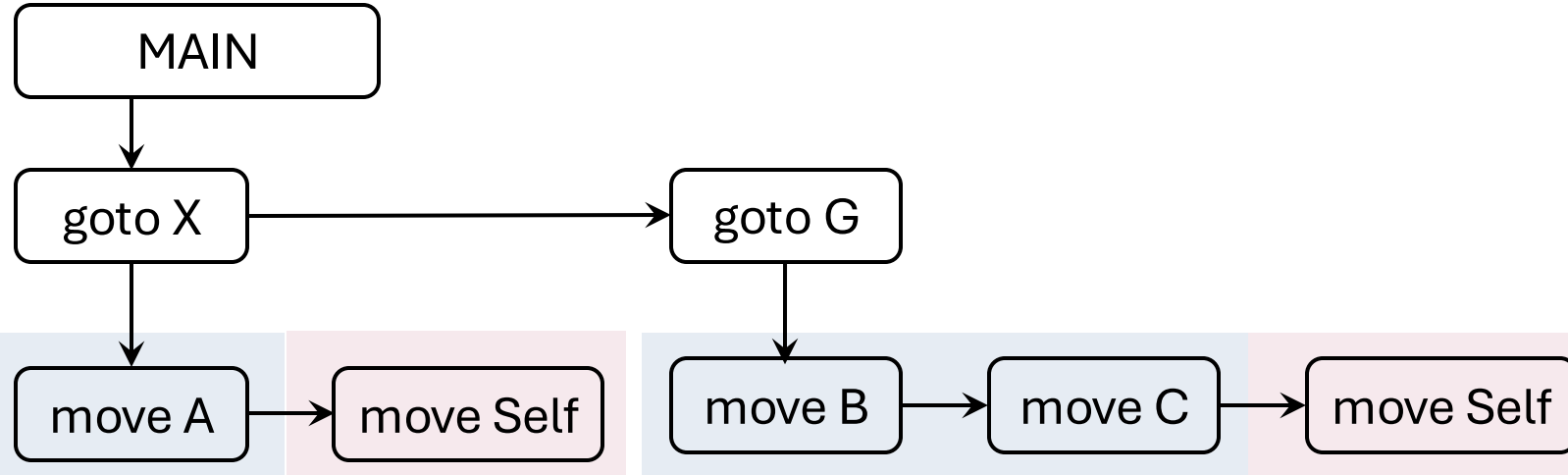
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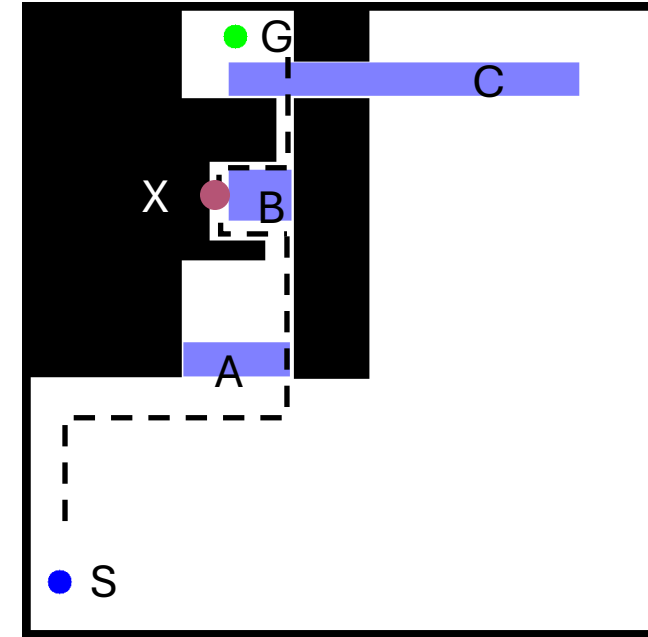
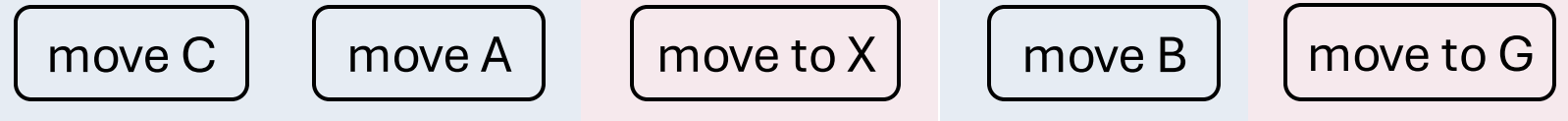
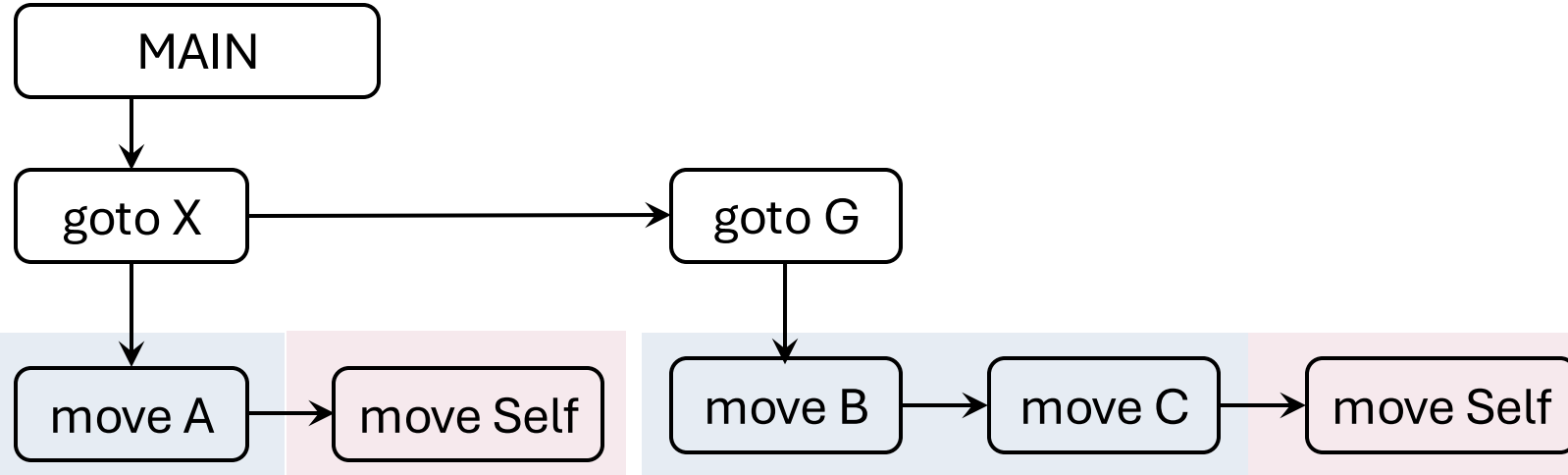
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+Promotion



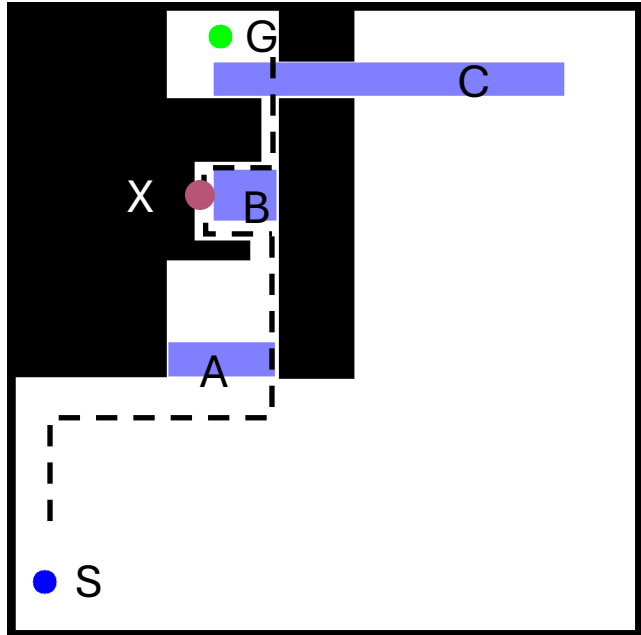
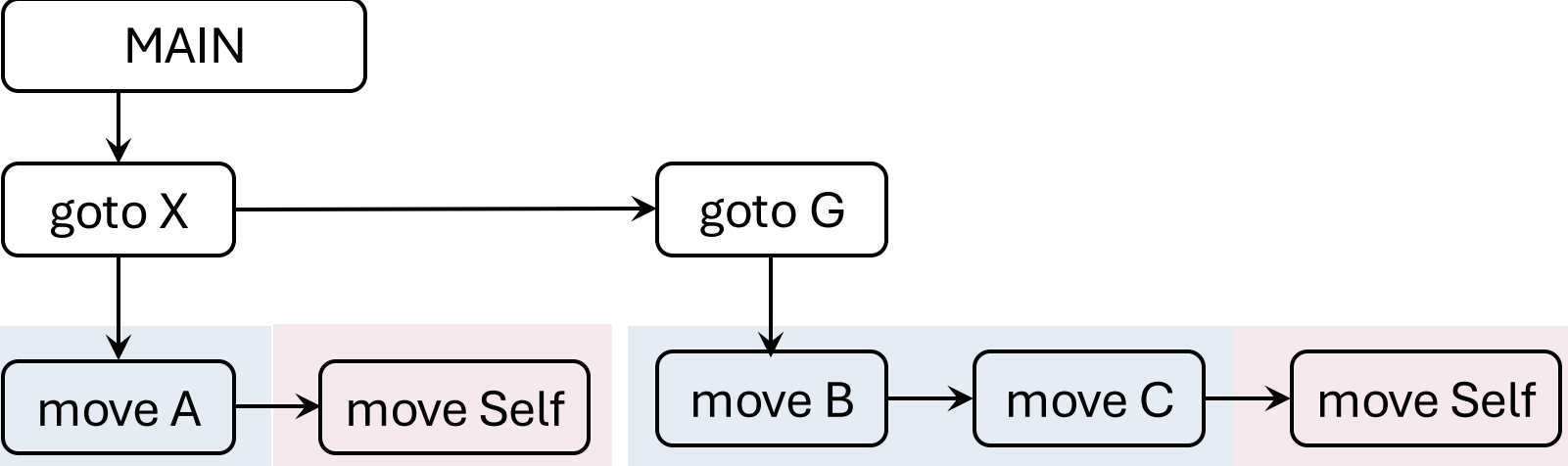
# Adding Flexible Promotion

Imperative

+Variables

+Ordering

+Promotion



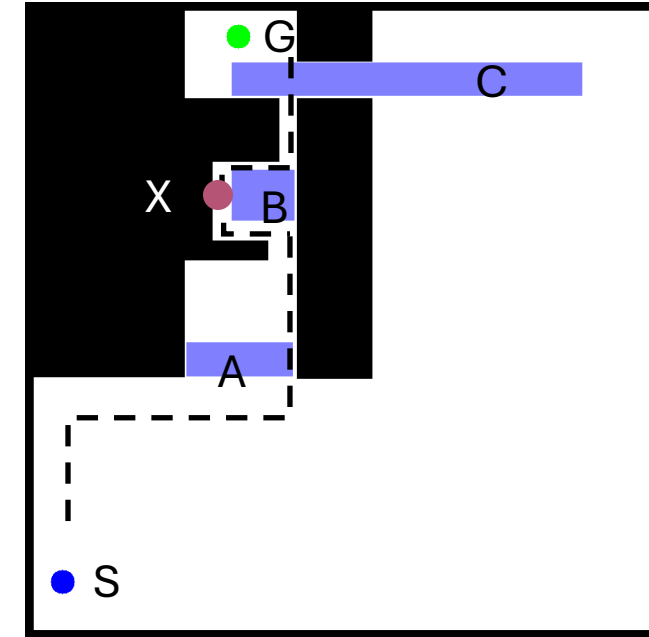
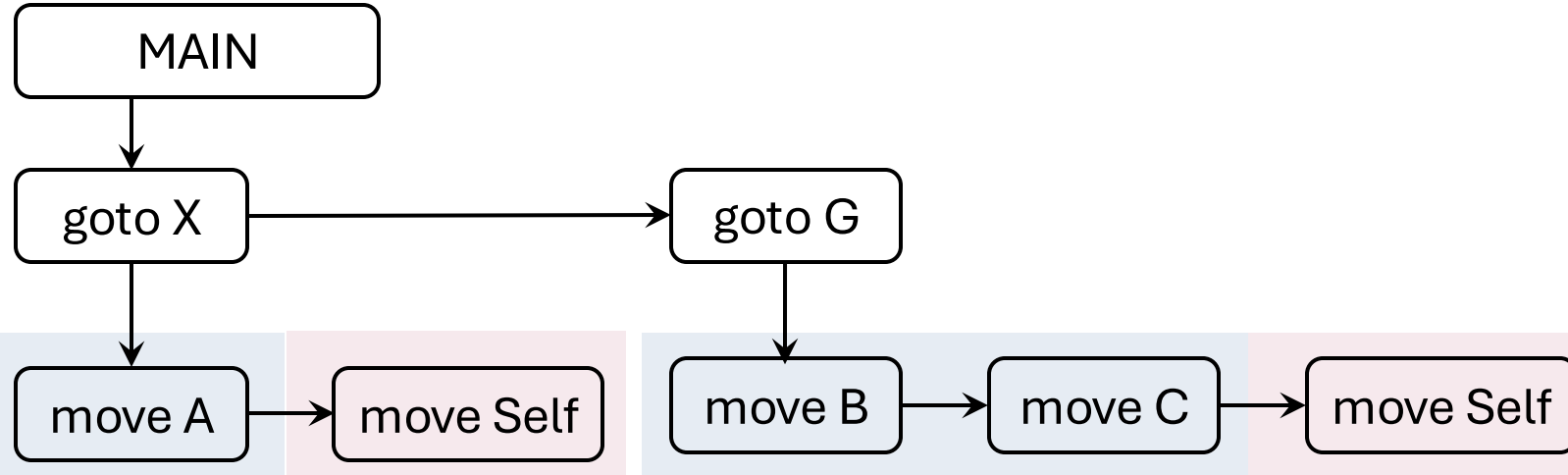
# Adding Flexible Promotion

Imperative

+Variables

+Ordering

+Promotion



# Adding Flexible Promotion

Imperative

+Variables

+Ordering

+Promotion

```
global_goal: agent_pos() == (270, 50)
```

```
behavior goto_v3(G: vector):
```

```
  goal: agent_pos() == G
```

```
  body:
```

```
    bind waypoint: vector
```

```
    achieve agent_pos() == waypoint
```

```
    bind path = find_path(agent_pos(), G)
```

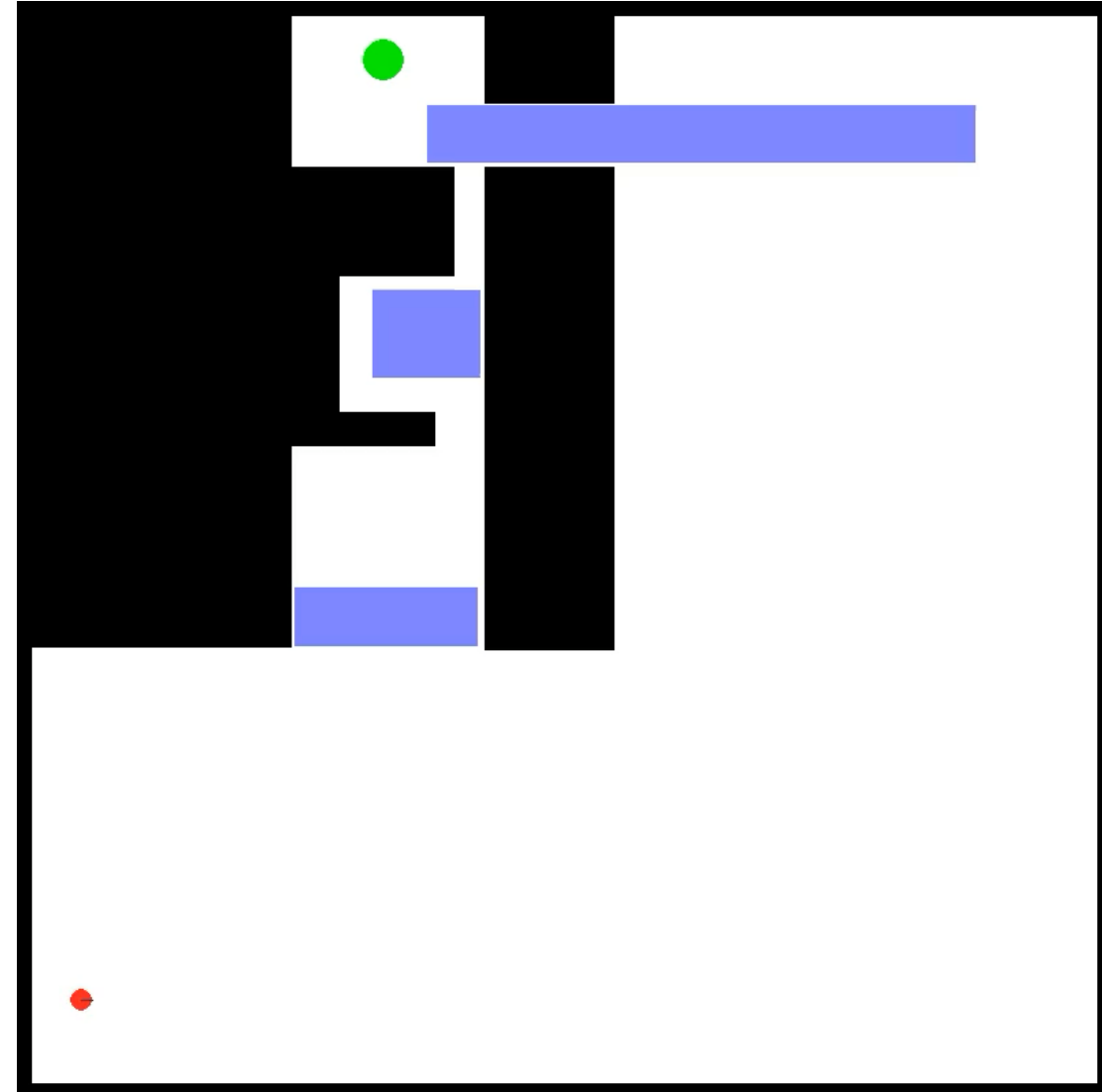
```
    promotable unordered:
```

```
      achieve not_blocking(A, path)
```

```
      achieve not_blocking(B, path)
```

```
      achieve not_blocking(C, path)
```

```
    do move_path(path)
```





# The Spectrum Between Imperative and Declarative



**Insight 1:** Behaviors = Generators of “non-deterministic subroutine calls”  
+ Verifies using causal models: pre- and post-conditions




**Insight 2:** Declarative = Imperative + Variable + Ordering + Promotion

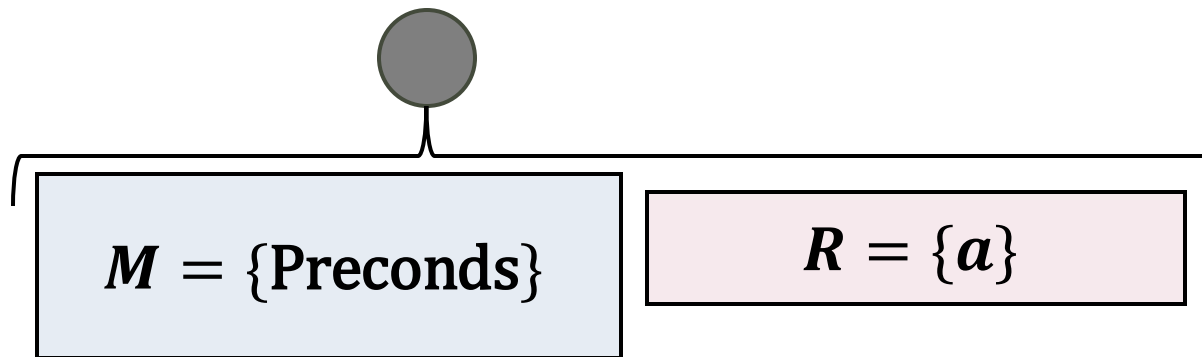
Specifically, if you only denote:

- the variables needed
- the preconditions they need to satisfy
- no ordering information about how preconditions should be achieved

You get full declarative modeling

# Reformulate Classical Formulations by “Language Feats.”

	+Variables Binding	+Ordering	+Promotion
PDDL	 Discrete Only	 Always	 Always



Do not support



Support, but you can't configure



Support, and configurable

# Reformulate Classical Formulations by “Language Feats.”

	+Variable Binding	+Ordering	+Promotion
PDDL	 Discrete Only	 Always	 Always
HTN/HGN	 Discrete Only		 Always
GOLOG	 Discrete Only		 Manual Interleaving
PDDLStream		 Always	 Always
<b>CDL</b> (Ours)			



Do not support









Support, but you can't configure



Support, and configurable

# Reformulate Classical Formulations by “Language Feats.”

	+Variable Binding	+Ordering	+Promotion
PDDL	 Discrete Only	 Always	 Always
HTN/HGN			
GOLOG			
PDDLStream			
<b>CDL (Ours)</b>			

## The Crow Planner:

Consumes the flexible representations

- Sound
- Probabilistically resolution complete



Do not support

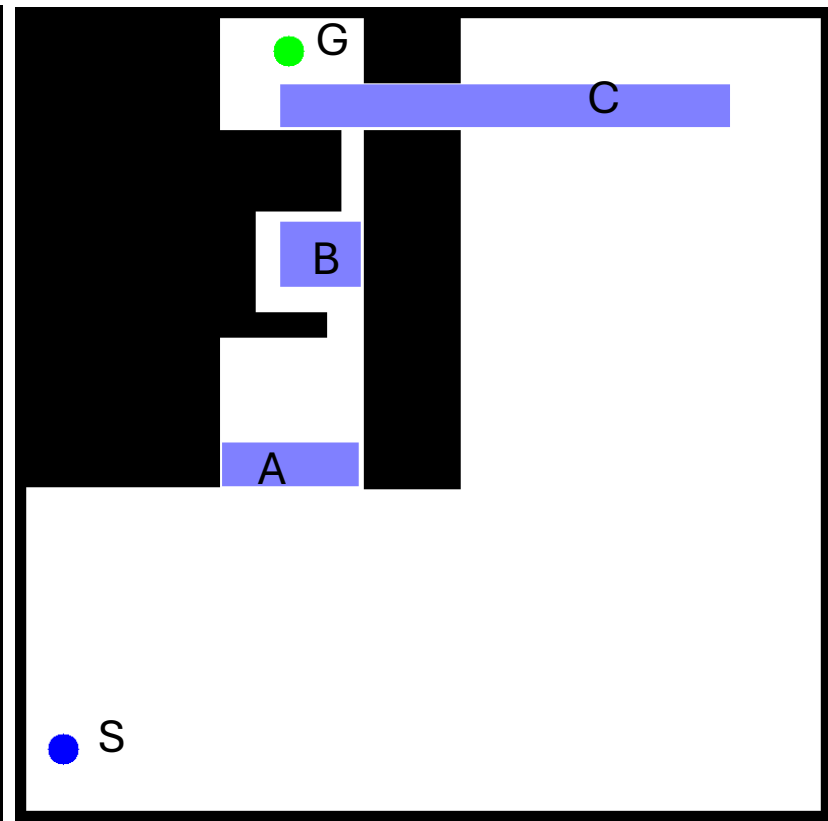
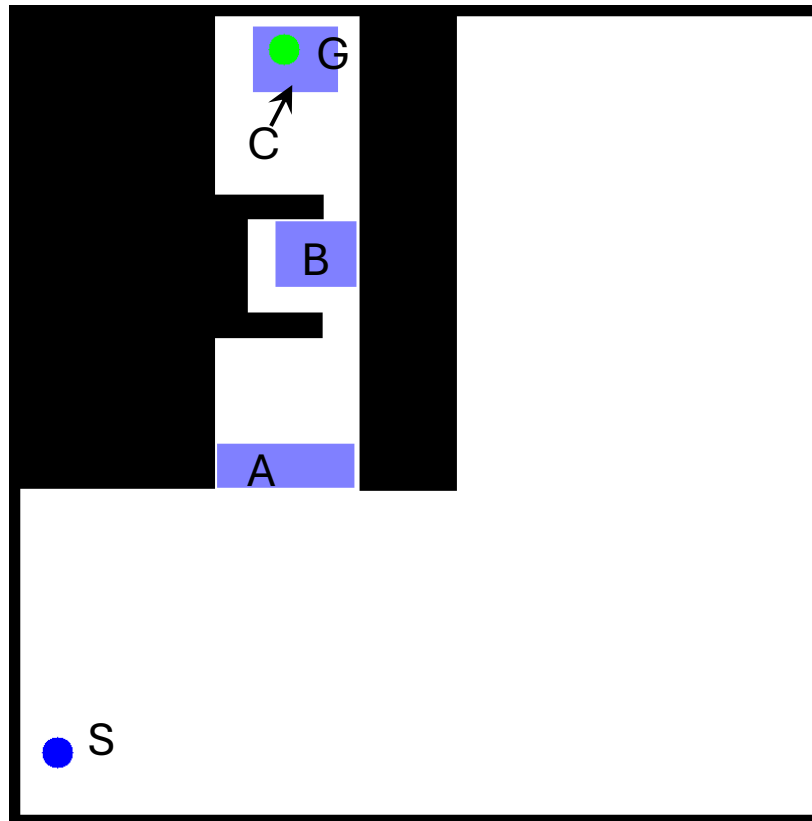
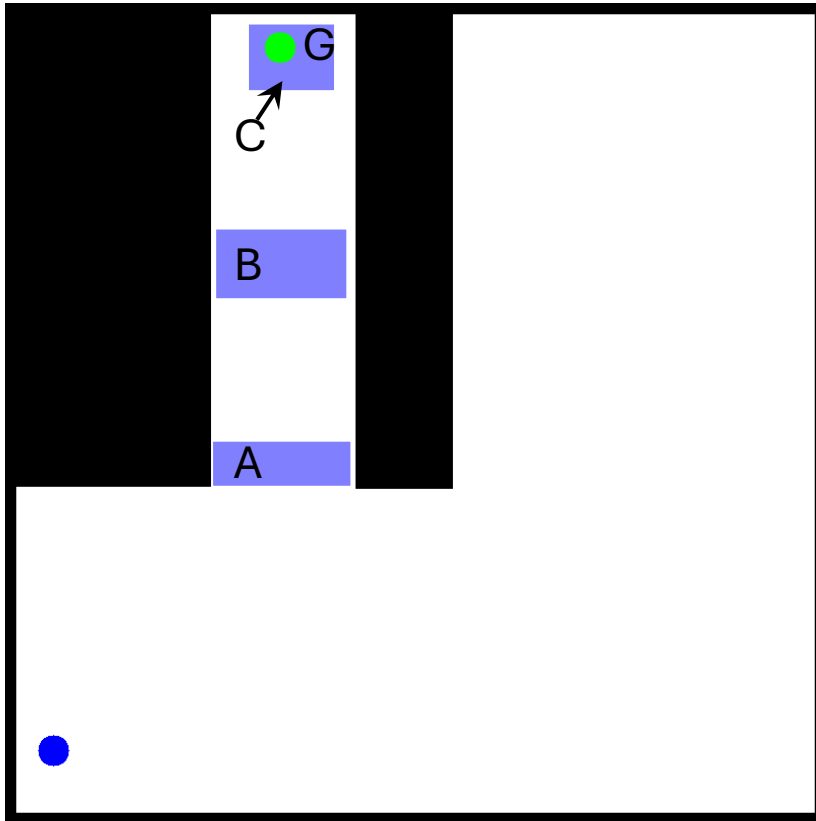


Support, but you can't configure



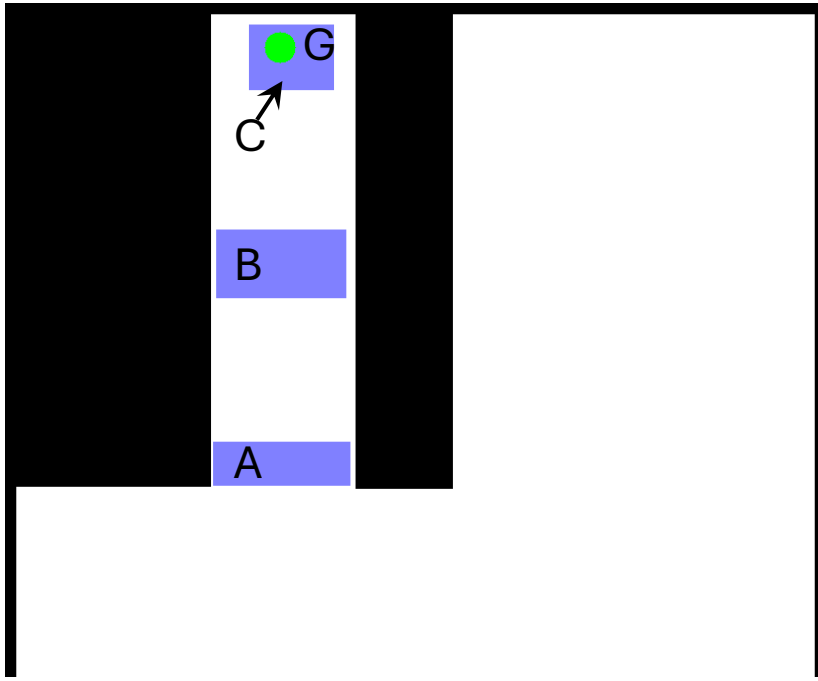
Support, and configurable

# Application: Context-Specific Solution Strategies

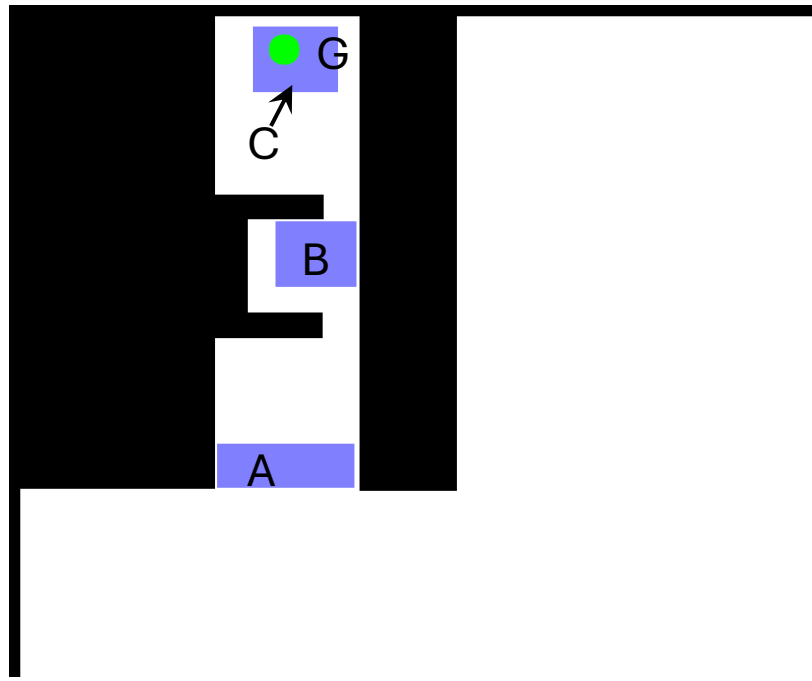


**Mid Level:** Closely related to the *LP1* class in Stilman and Kuffner 2005, “disconnected spaces can be connected by moving a single obstacle”

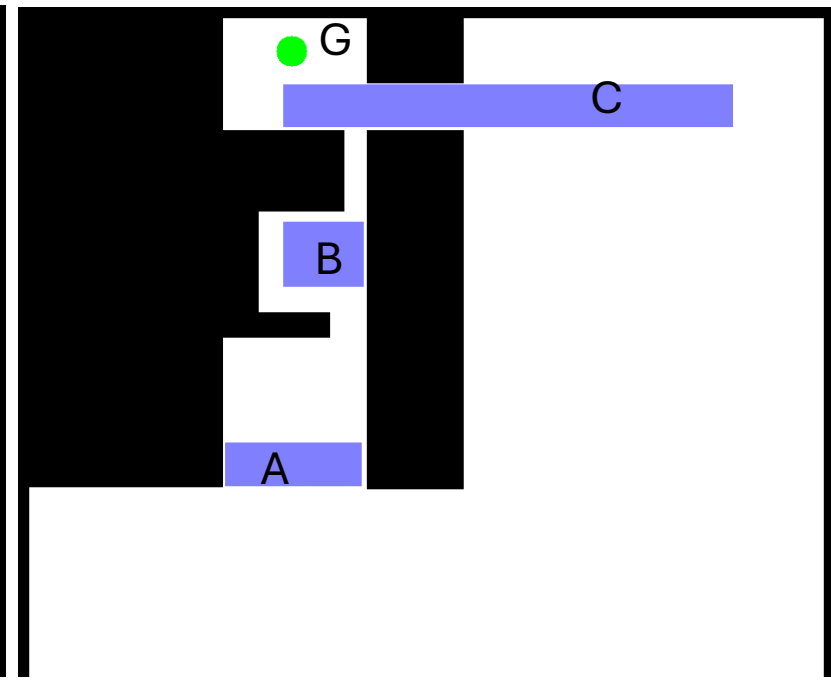
# Application: Context-Specific Solution Strategies



```
behavior goto_v1(G: vector):  
  goal: agent_pos() == G  
  body:  
    bind path = find_path(agent_pos(), G)  
    unordered:  
      achieve not_blocking(A, path)  
      achieve not_blocking(B, path)  
      achieve not_blocking(C, path)  
    do move_path(path)
```

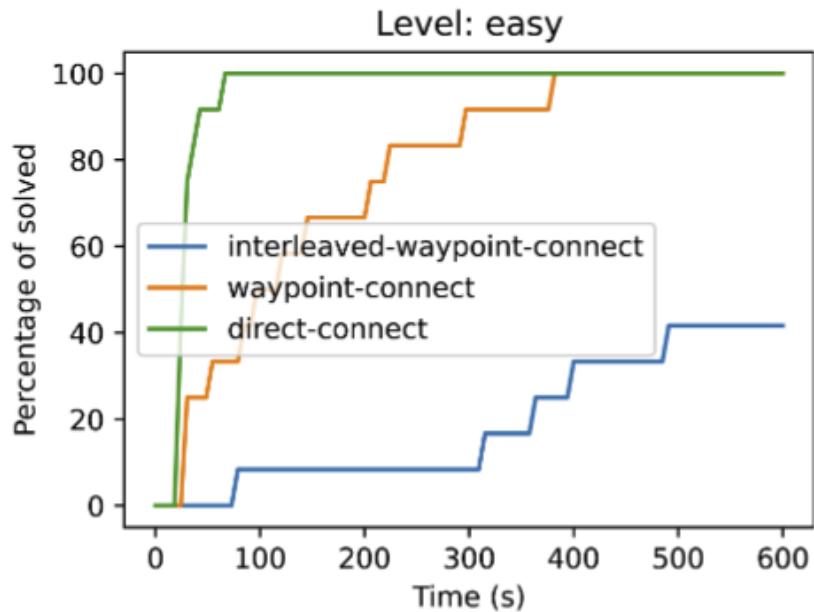
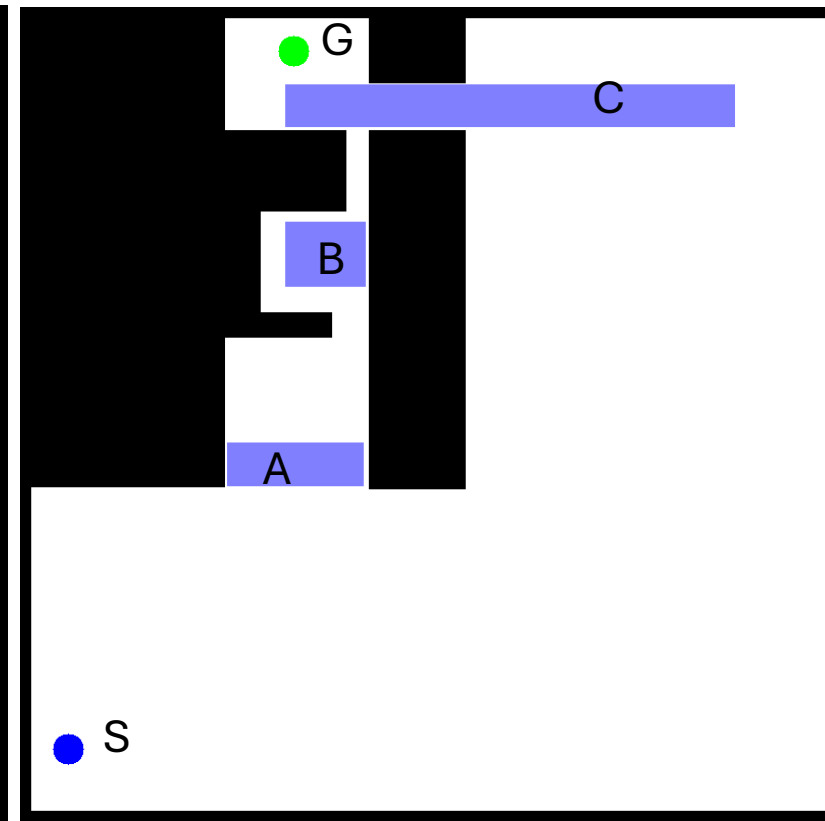
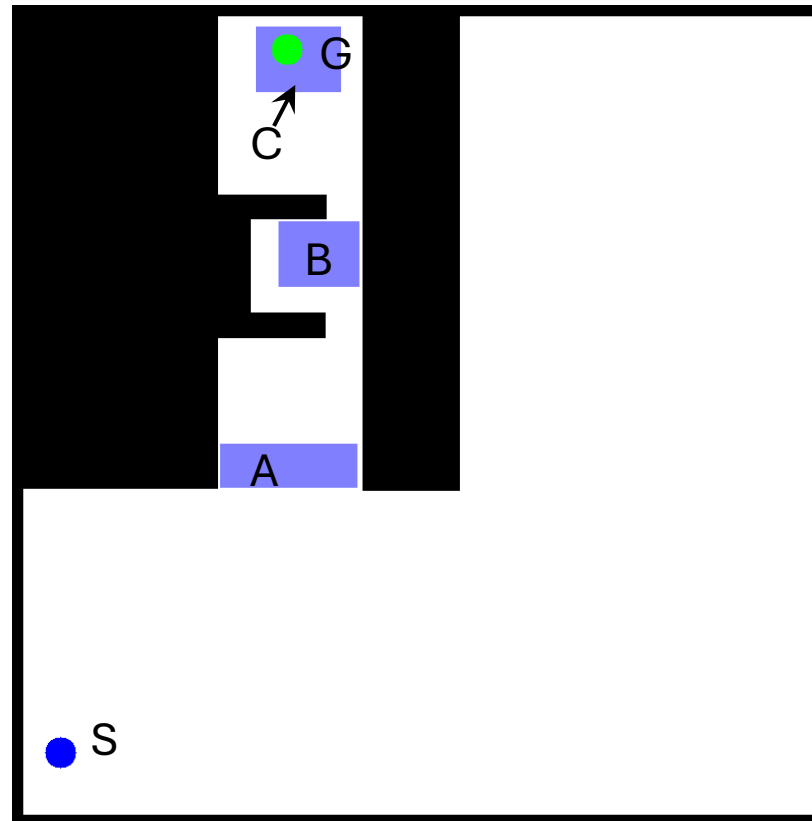
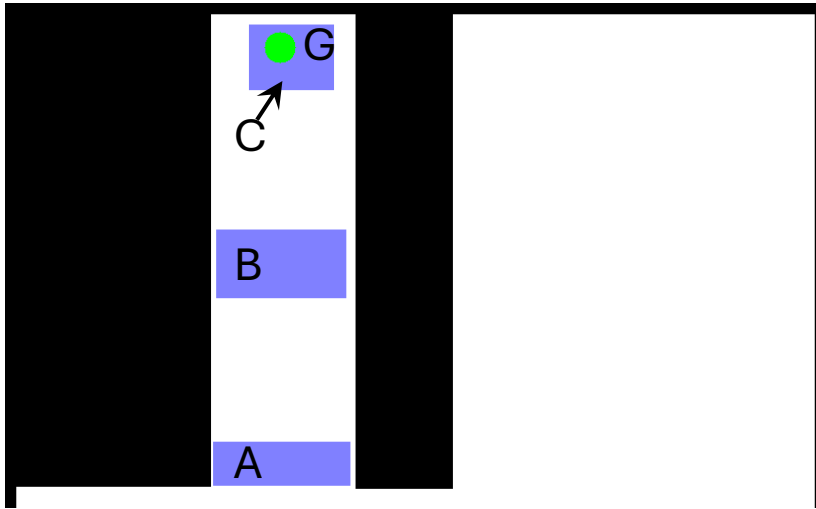


```
behavior goto_v2(G: vector):  
  goal: agent_pos() == G  
  body:  
    bind waypoint: vector  
    achieve agent_pos() == waypoint  
    bind path = find_path(agent_pos(), G)  
    unordered:  
      achieve not_blocking(A, path)  
      achieve not_blocking(B, path)  
      achieve not_blocking(C, path)  
    do move_path(path)
```

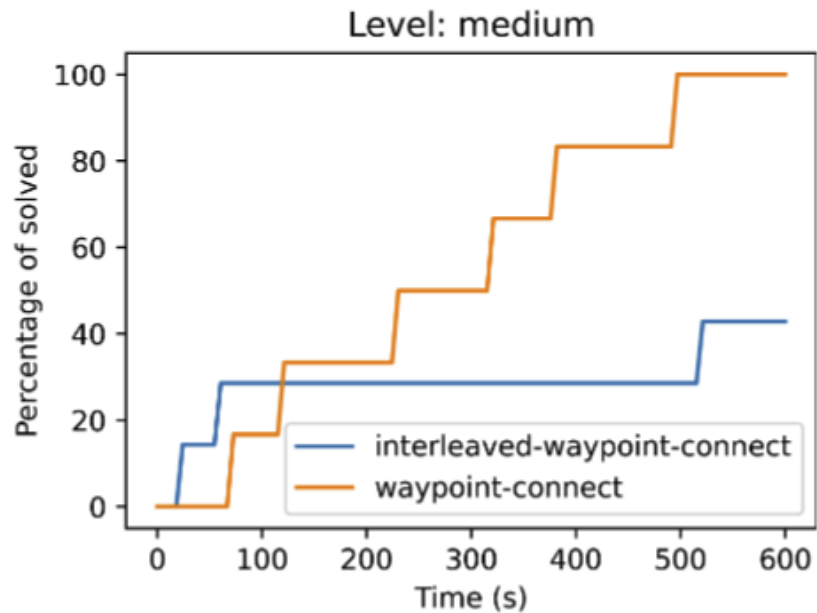
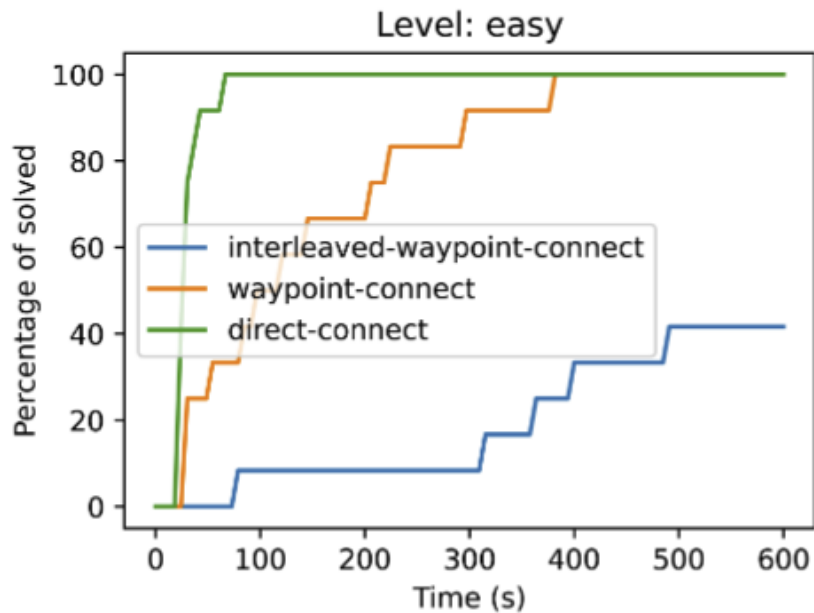
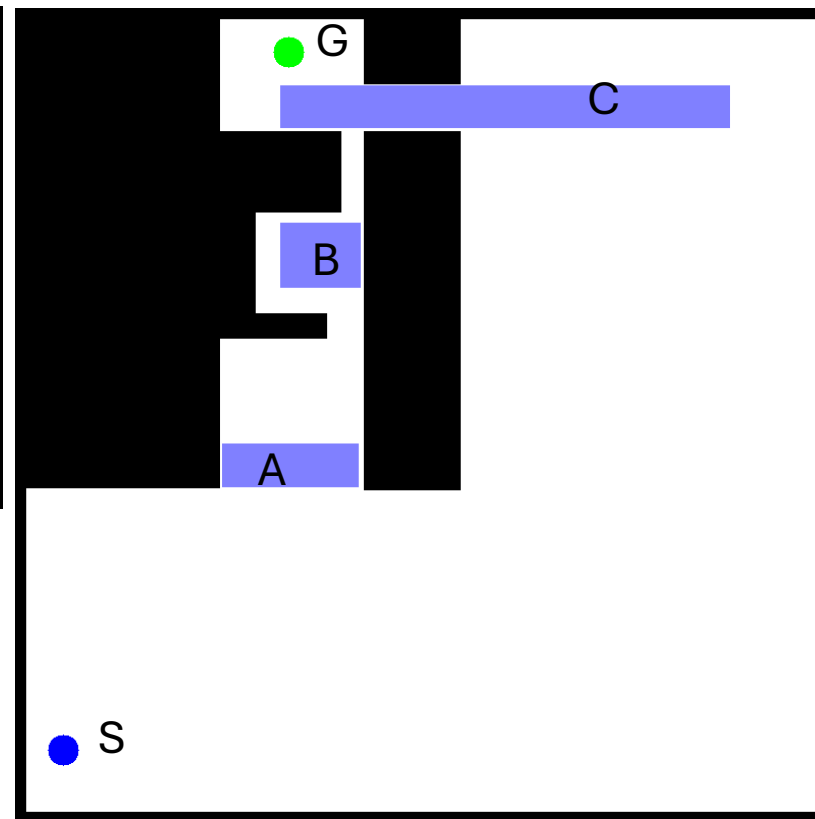
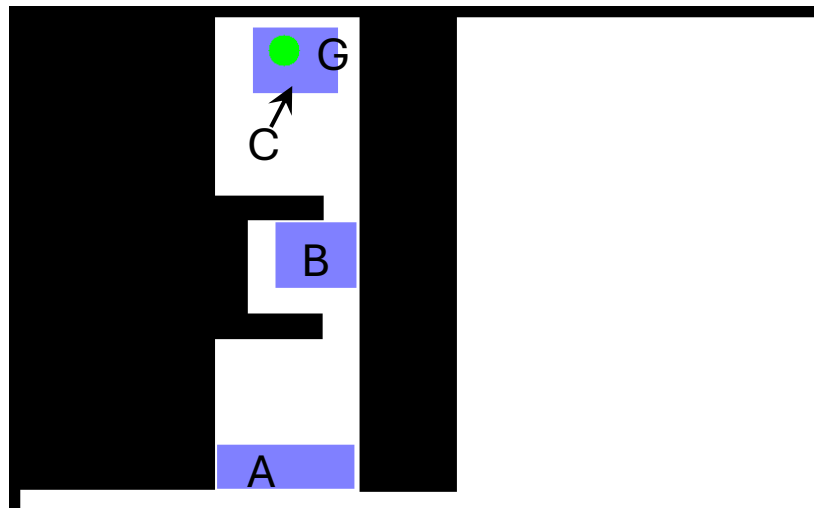
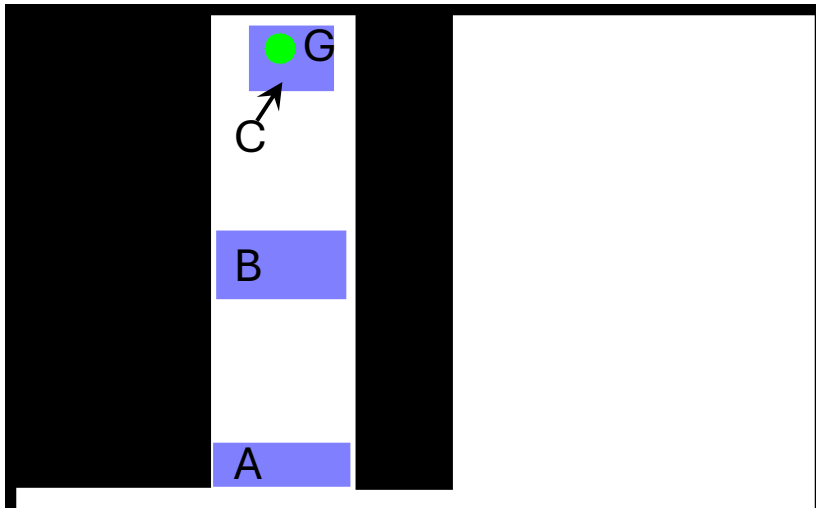


```
behavior goto_v3(G: vector):  
  goal: agent_pos() == G  
  body:  
    bind waypoint: vector  
    achieve agent_pos() == waypoint  
    bind path = find_path(agent_pos(), G)  
    promotable unordered:  
      achieve not_blocking(A, path)  
      achieve not_blocking(B, path)  
      achieve not_blocking(C, path)  
    do move_path(path)
```

# Context-Specific Strategies Improves Efficiency

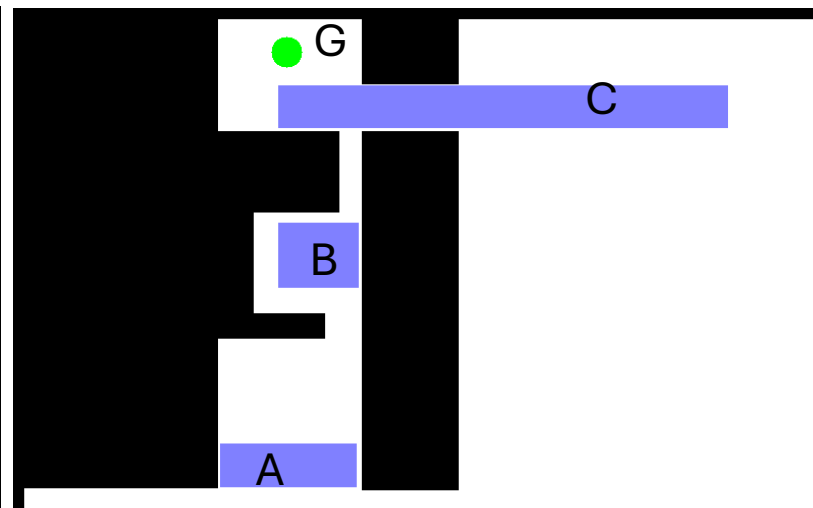
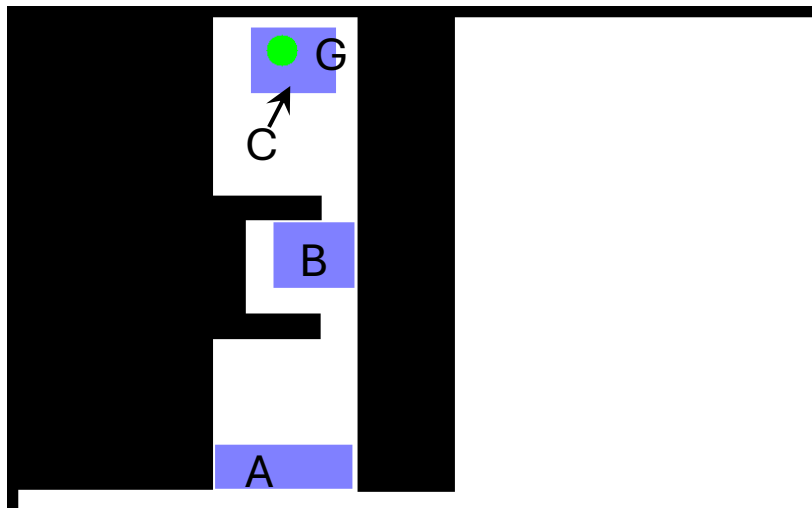
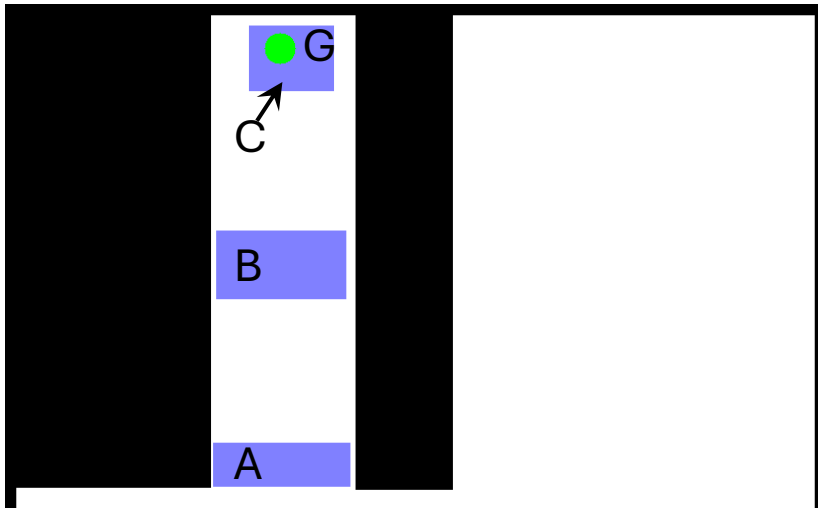


# Context-Specific Strategies Improves Efficiency

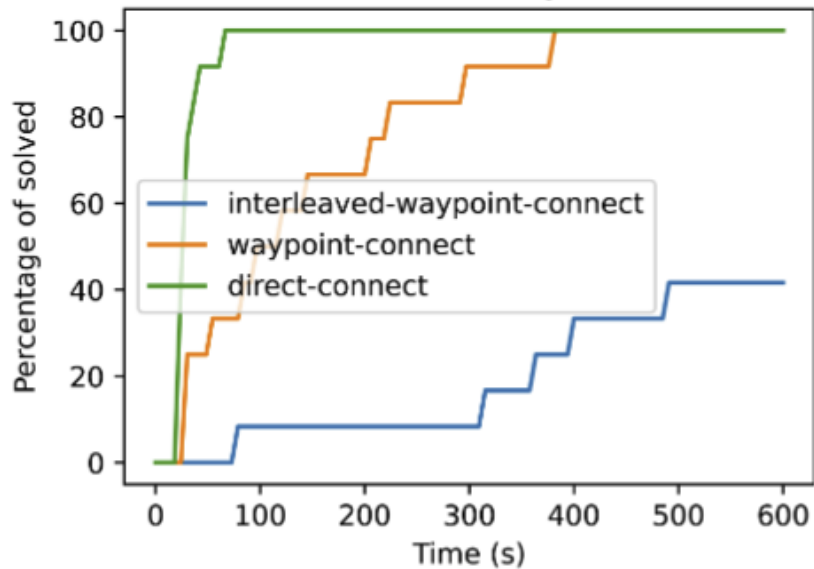




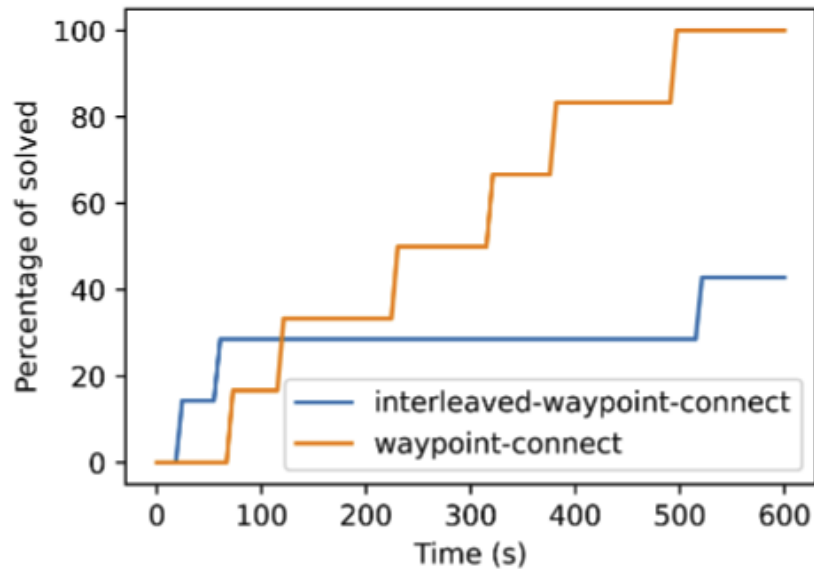
# Context-Specific Strategies Improves Efficiency



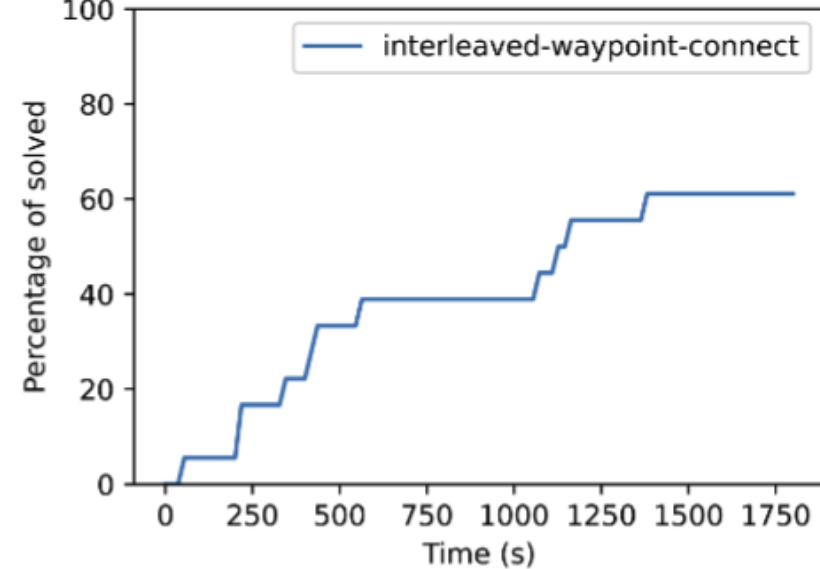
Level: easy



Level: medium



Level: hard



# Theory: Planning Complexity of Problems

Promotable  
Section  $M$

Serialized Section  $R$

**Theorem (very informally):** under serializability assumptions over  $R$ , the planning complexity is bounded by  $n^{O(k)}$ , where  $n$  is the number of objects,  $k$  is the maximum number of subgoals that would accumulate in  $M$

**Intuition:**  $k$  defines how easy it is to “serialize” a problem

- **NAMO:**  $k$  is the number of obstacles that have “dependencies”

Closely Related to “Width” in Symbolic Planning and Neural Network Expressivity

Lipovetzky and Geffner. 2012. “Width and serialization of classical planning problems”

Mao et al. 2023. “What Planning Problem Can A Relational Neural Network Solve?”

# Dirty Laundry

## Theory

- The bound is not tight because it treats all objects “uniformly”
- Ultimately, what we really want to is to identify the “kernel” of the problem

## Practice

- Although we support description of different solution strategies compactly,
- we do not know which one to apply
- Actually, this can be as hard as solving the original problem

# Conclusion



**Principle:** Using program semantics to characterize flexibilities in behaviors

We provide a new framework for “how to plan more efficiently”

- **Theory:** characterize the hardness of a problem
- **Practice:** a framework for mix-and-match representations

**Next:** learning how to reason more efficiently

- learning to select the best strategy in context
- learning to form new strategies, by reasoning about different types of flexibilities