

CS 460/560

Introduction to Computational Robotics
Fall 2019, Rutgers University

Lecture 19

More Planning Problems and Methods

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Outline

Time varying motion planning problems

Mixing discrete and continuous spaces

Coverage planning

Sensor-based planning methods

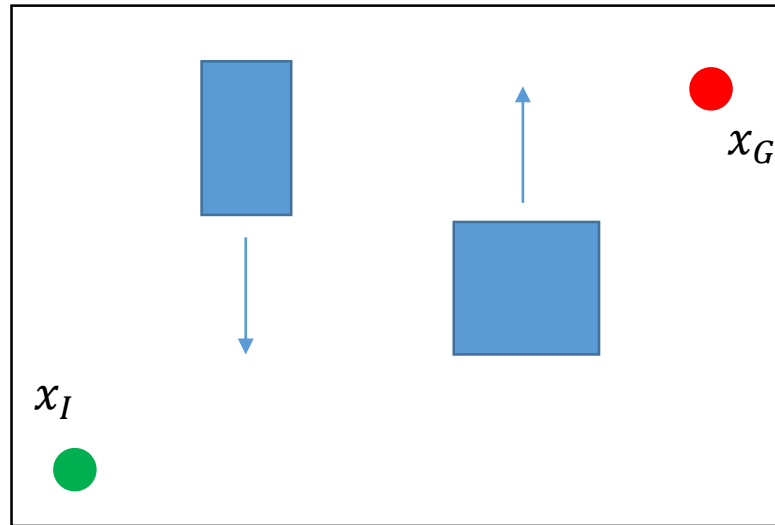
- ⇒ Bug algorithms

- ⇒ Gap-navigation trees

Time Varying Problems

So far we have only discussed **static** problems

- ⇒ That is, the obstacles are static
- ⇒ Also known as **time-invariant** problems
- ⇒ When obstacles may change over time, the problem is **time varying**
- ⇒ E.g.

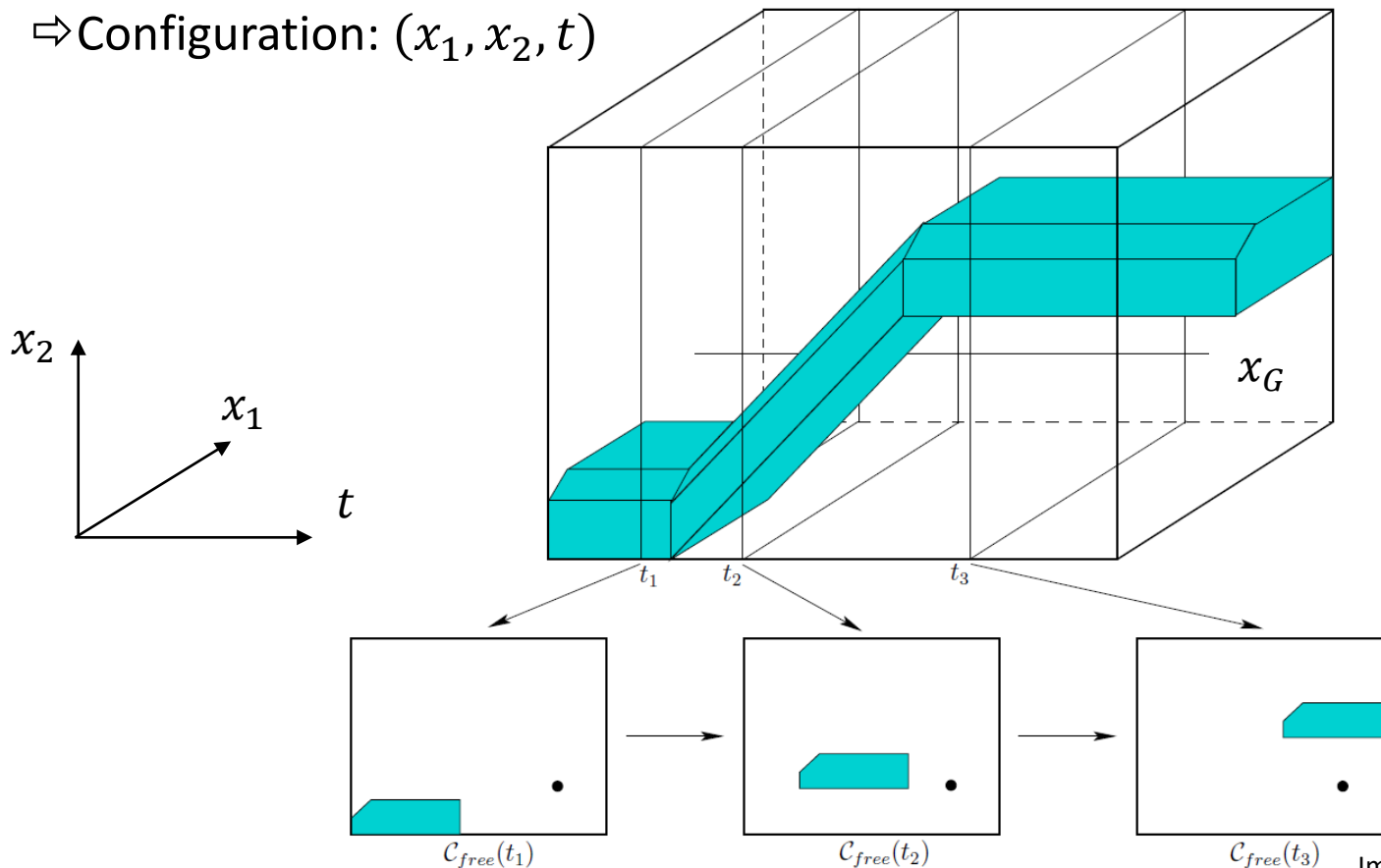


- ⇒ One may view multi-robot problems as a time varying problem as well

Predictable Obstacles

When obstacle movement is deterministic, the problem can be turned into a problem that looks like a time-invariant problem

- ⇒ This is done via adding the extra “time” dimension
- ⇒ E.g., single obstacle, piece wise linear motion
- ⇒ Configuration: (x_1, x_2, t)



Solution with Unbounded Speed (I)

First consider problems without a “speed limit”

⇒ How can we solve the problem under such assumptions?

⇒ Combinatorial methods

⇒ E.g., vertical cell decomposition

⇒ Key: no traveling back in time!

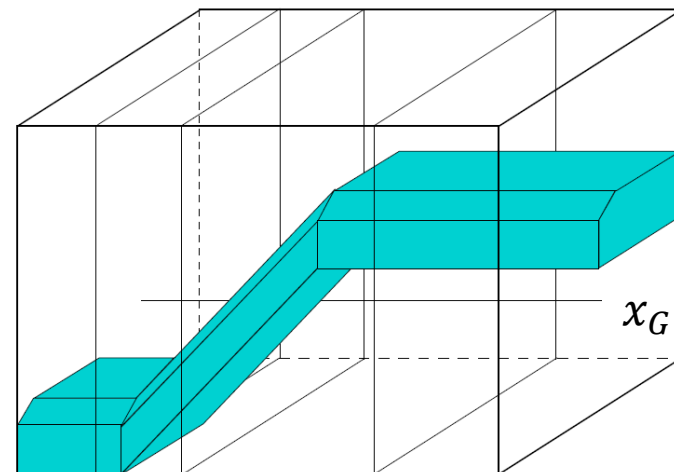
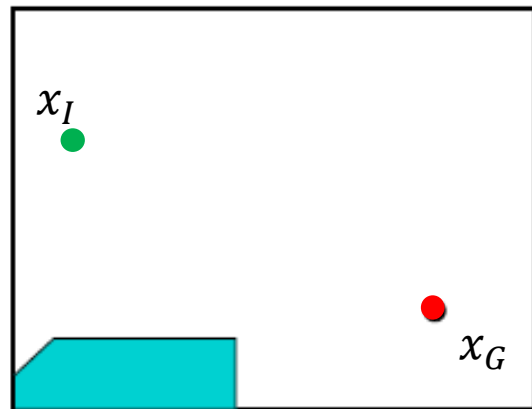
⇒ Probabilistic methods

⇒ PRM, RRT

⇒ Need a new metric: $d((x_1, x_2, t), (x'_1, x'_2, t')) = \begin{cases} 0 & \text{if } (x_1, x_2, t) = (x'_1, x'_2, t') \\ \infty & \text{if } (x_1, x_2, t) \neq (x'_1, x'_2, t') \text{ and } t' \leq t \\ d((x_1, x_2), (x'_1, x'_2)) & \text{otherwise} \end{cases}$

⇒ This is a **pseudo-metric**: not symmetric

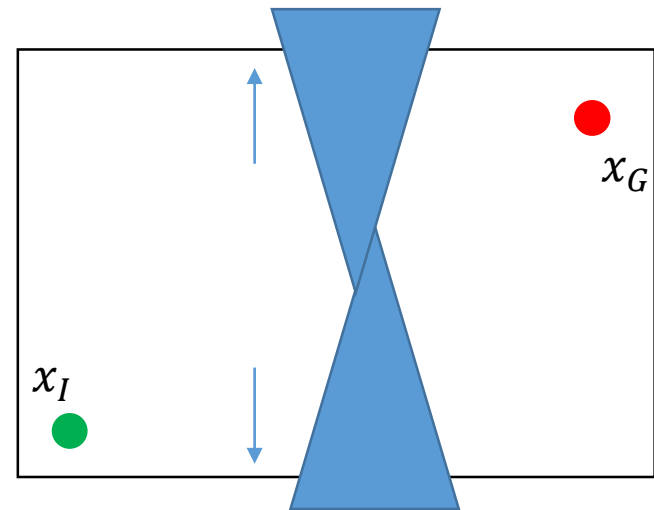
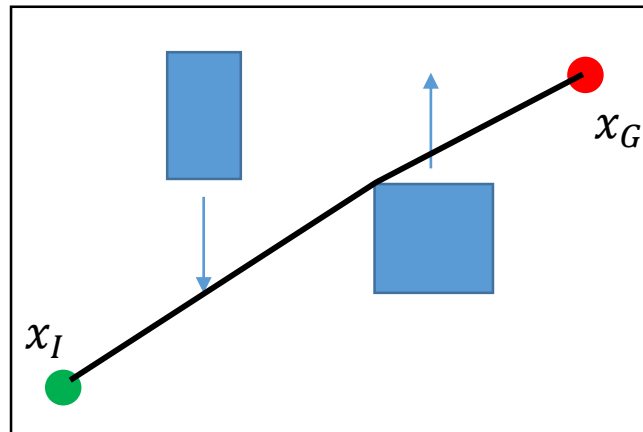
⇒ Basically, cannot travel back in time



Solution with Unbounded Speed (II)

Question: given unbounded speed, why not

- ⇒ Plan a path at $t = 0$
- ⇒ Ask the robot to just go fast to follow the path?
- ⇒ Because it does not always work!
 - ⇒ E.g.



Solution with Bounded Speed

For real problems, cannot use arbitrarily fast travel!

⇒ Otherwise, solution becomes “trivial” if feasible path exists at $t = 0$

⇒ But, we do not need to much more to make it work

Combinatorial methods

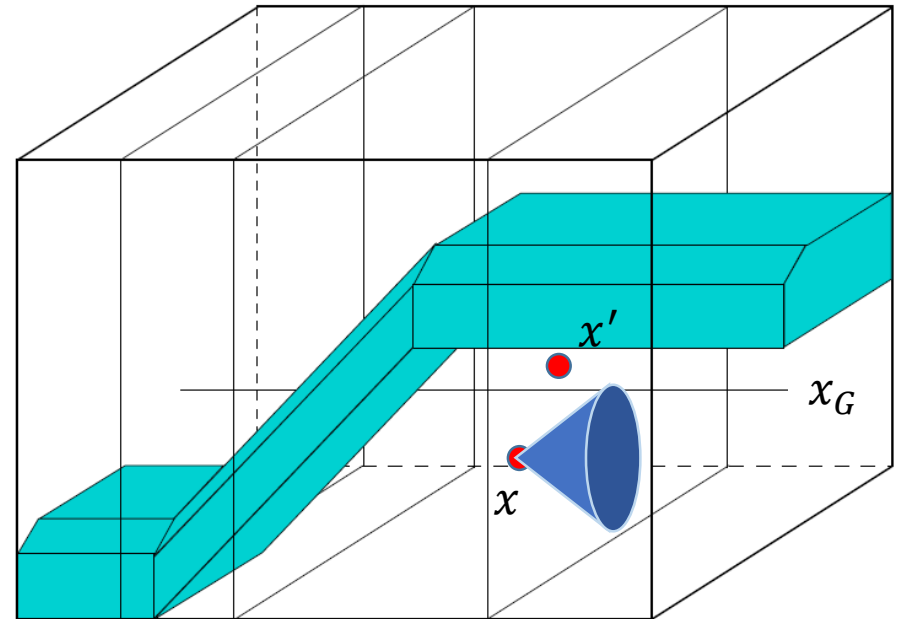
⇒ Ensure that the slope of any path segment has bounded magnitude

⇒ For any point in the space-time space, this induces a “speed cone”

Sampling based methods

⇒ Need to modify the pseudometric

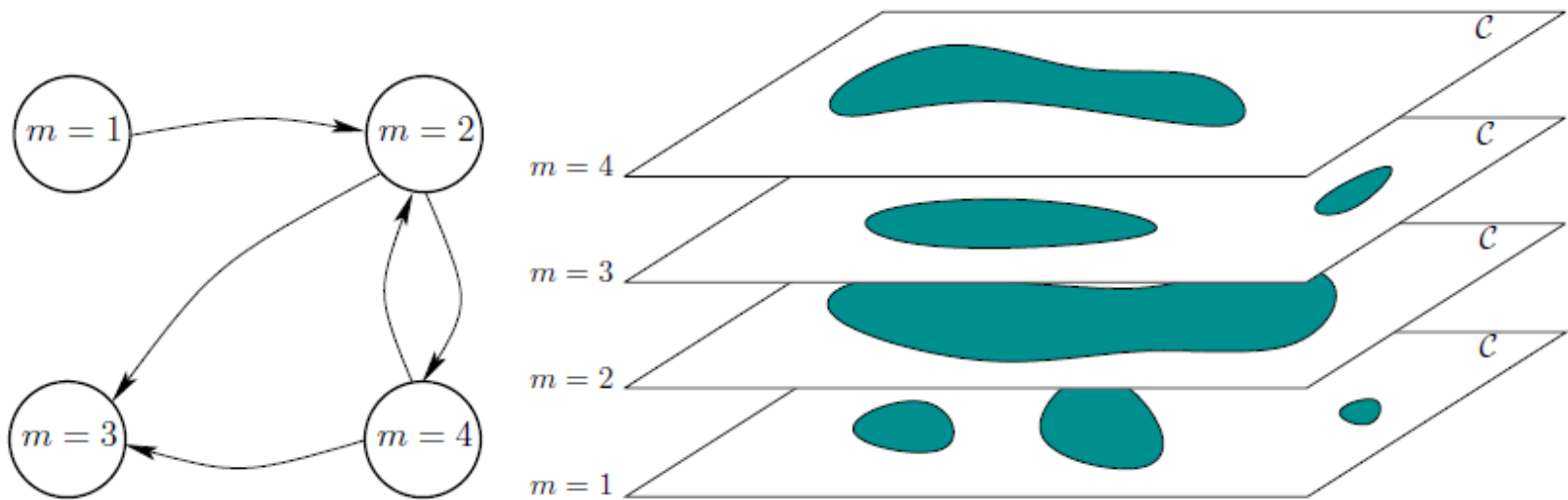
⇒ Same idea



Hybrid Systems

Hybrid systems is a model for mixing discrete and continuous domains

- ⇒ The system generally works with continuous domains
- ⇒ But it is also capable of making discrete transitions
- ⇒ In this case, the different continuous domains are “modes”
- ⇒ For example, the transmission system of a car



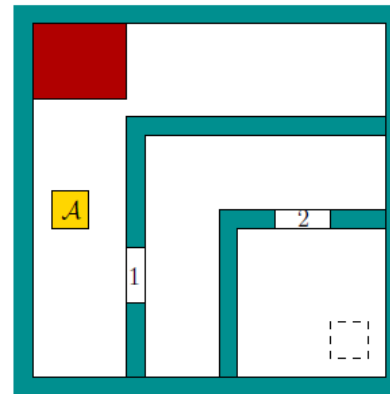
Hybrid System – A 2D Planning Example

The power of the portiernia

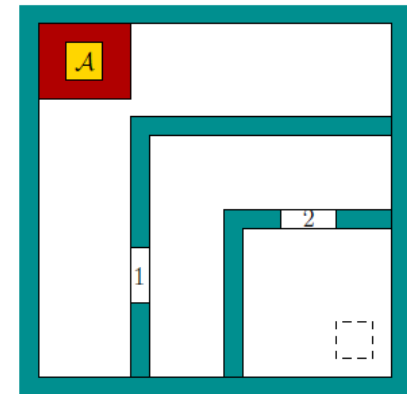
- ⇒ Robot must open door 1 and door 2 to access its goal
- ⇒ It may retrieve one key at a time from the portiernia
- ⇒ Discrete elements:
 - ⇒ Which key the robot has
 - ⇒ The status of the door
- ⇒ Continuous elements:
 - ⇒ The 2D domain that may change

Calls for **task and motion planning**

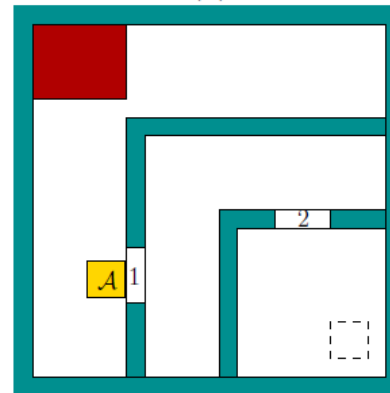
- ⇒ Can be modeled as special automata
 - ⇒ Can do planning and verification
- ⇒ Or, encode with first order logic
 - ⇒ E.g., planning domain definition language



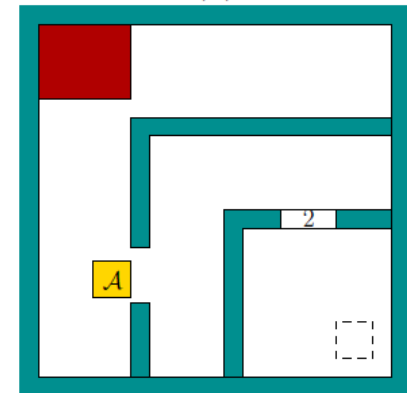
(a)



(b)



(c)

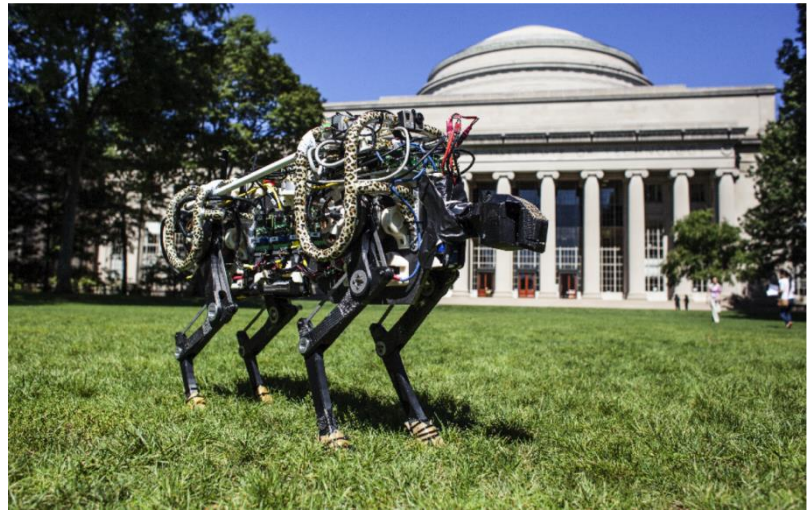


(d)

Configuration Space with Varying Dimensions

Another scenario that mixes discrete and continuous spaces has changing configuration space dimensions

- ⇒ Bipedal robots, legged robots
- ⇒ Why?
- ⇒ Manipulators grasping/pushing objects
- ⇒ More on these later...



Coverage Planning (I)

An important problem in practice is **coverage planning**

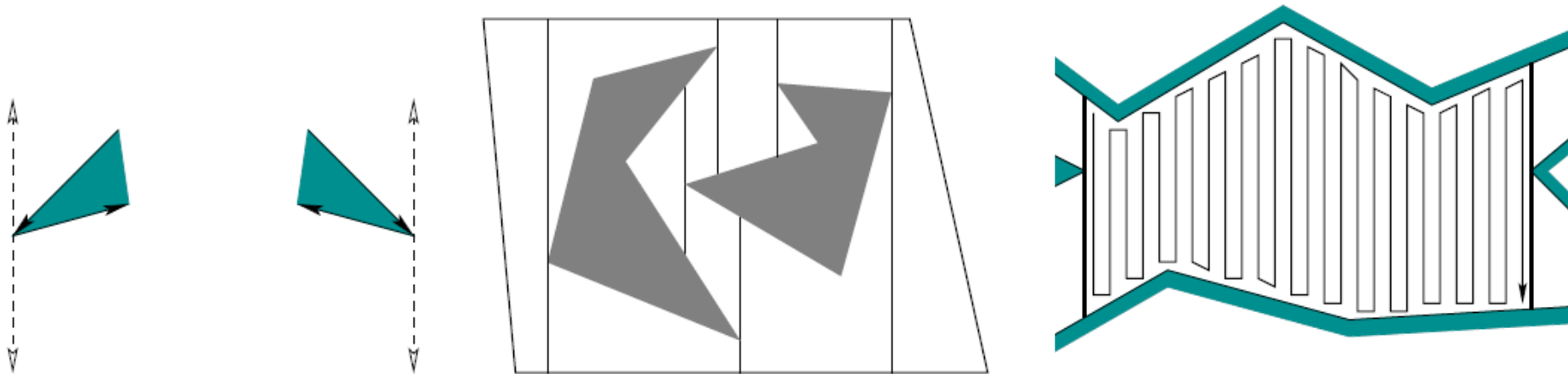
- ⇒ Planning paths for Roomba to clean a room
- ⇒ Planning paths for harvesting crops
- ⇒ What is the main goal here?
 - ⇒ Cover (using limited footprint) the space with least amount of total travel
 - ⇒ A second objective can be minimizing turns, which takes time



Coverage Planning (II)

Boustrophedon decomposition

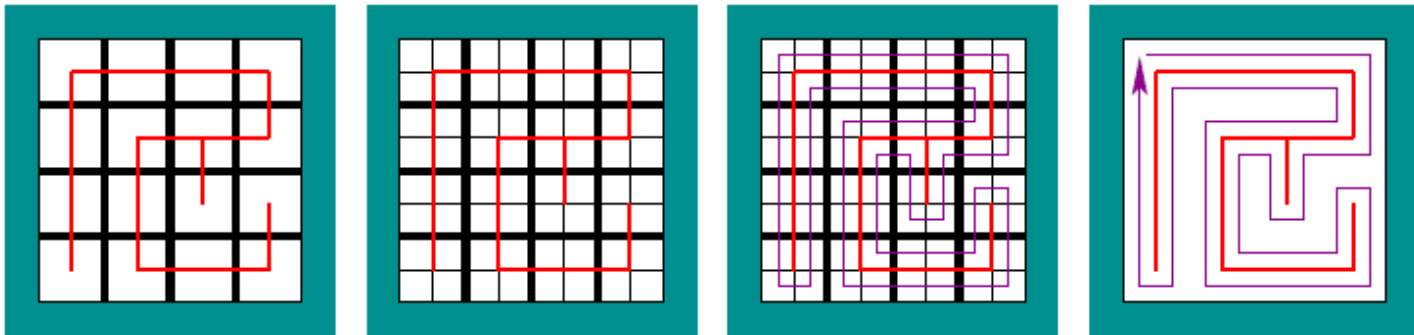
- ⇒ One way to do coverage planning is intuitive: move back and forth!
- ⇒ Basically, do a (special) vertical cell decomposition
- ⇒ Then, do boustrophedon paths in each cell
- ⇒ Issue: when cells split, may need to travel back and waste time
- ⇒ Can be resolved using spanning tree double covering



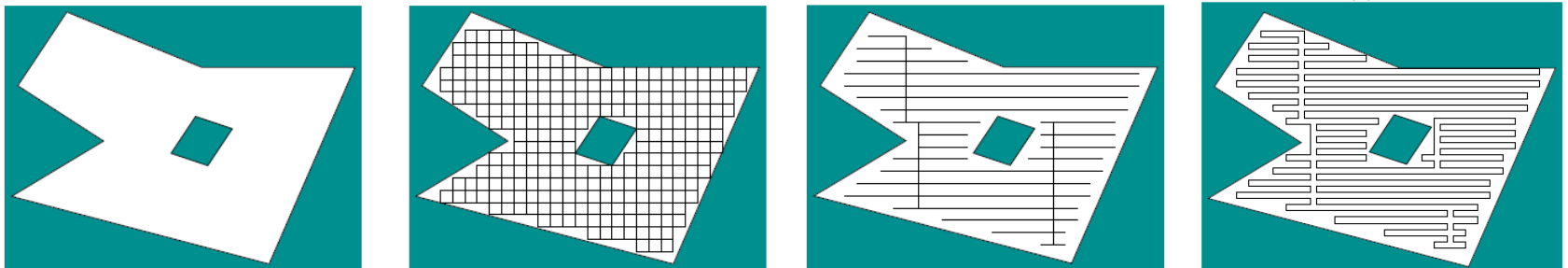
Coverage Planning (III)

Spanning tree double cover

- ⇒ Cover the free space with a grid
- ⇒ Find a spanning tree
- ⇒ Doubling up the spanning tree to create a single non-overlapping path



- ⇒ Example with obstacles



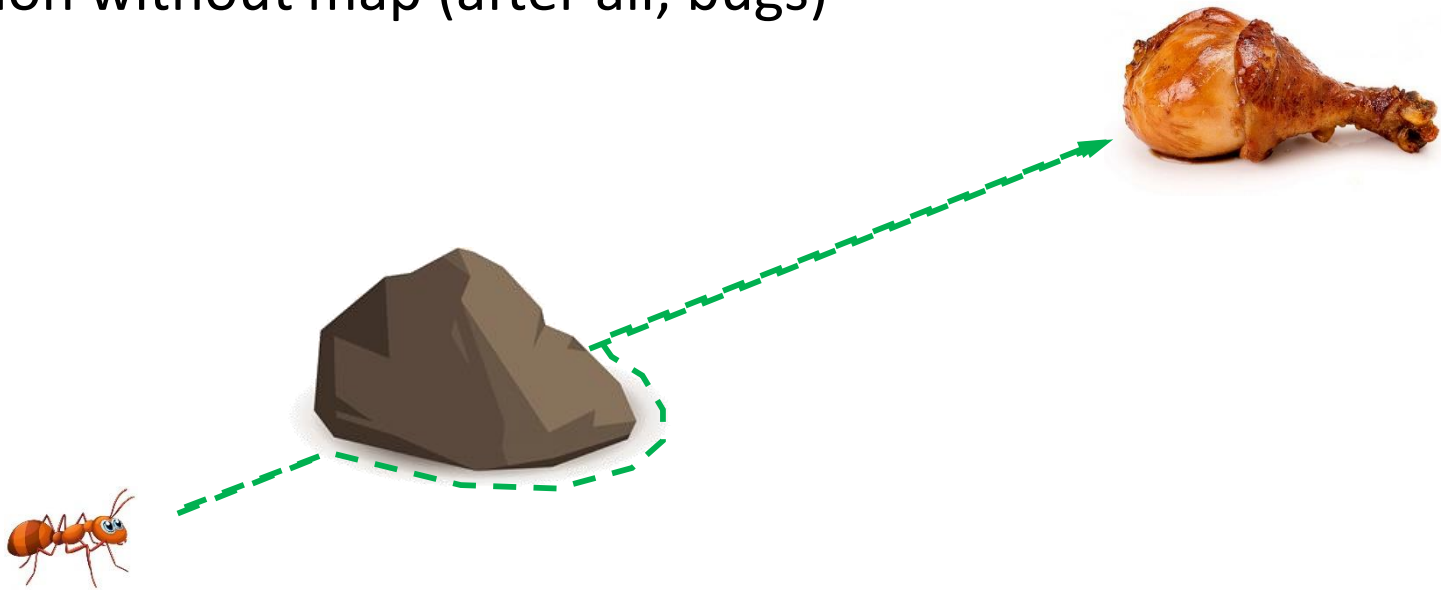
Bug Algorithms

Sometimes robots may have special sensors

For a bug, e.g., ants, the sensors may be

- ⇒ A goal sensor: detects the direction of the goal (e.g., food)
- ⇒ An obstacle sensor
- ⇒ Markers: the bug knows when it revisits a place

Navigation without map (after all, bugs)

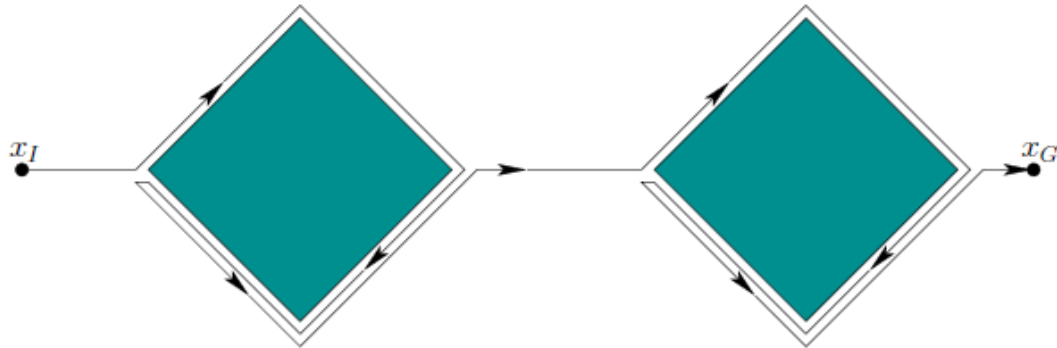


Bug1

The Bug1 algorithm: scout each obstacle

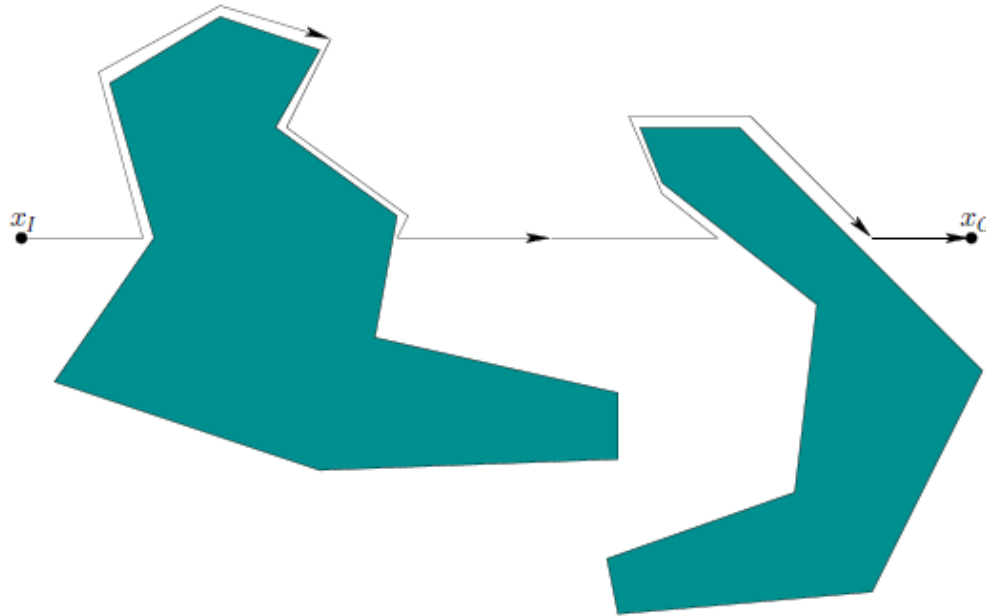


Complete but can be inefficient, e.g.,

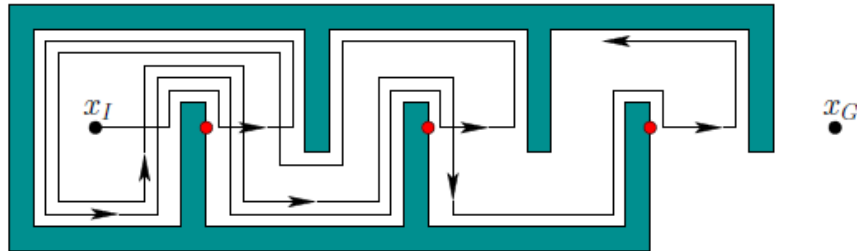


Bug2

Bug2 algorithm: work with a fixed bearing

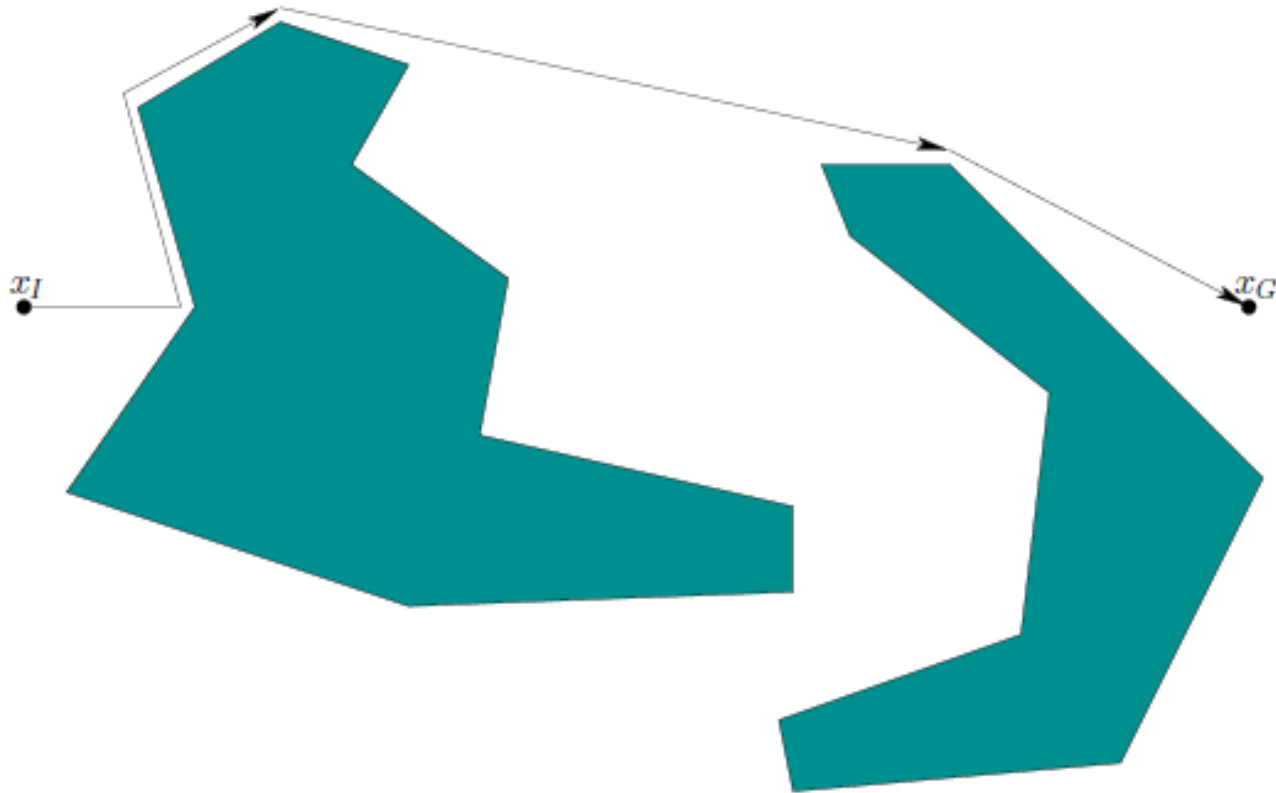


A bad case for Bug2



Others Bug Algorithms

Many other bug algorithms: VisBug, TangentBug, ...

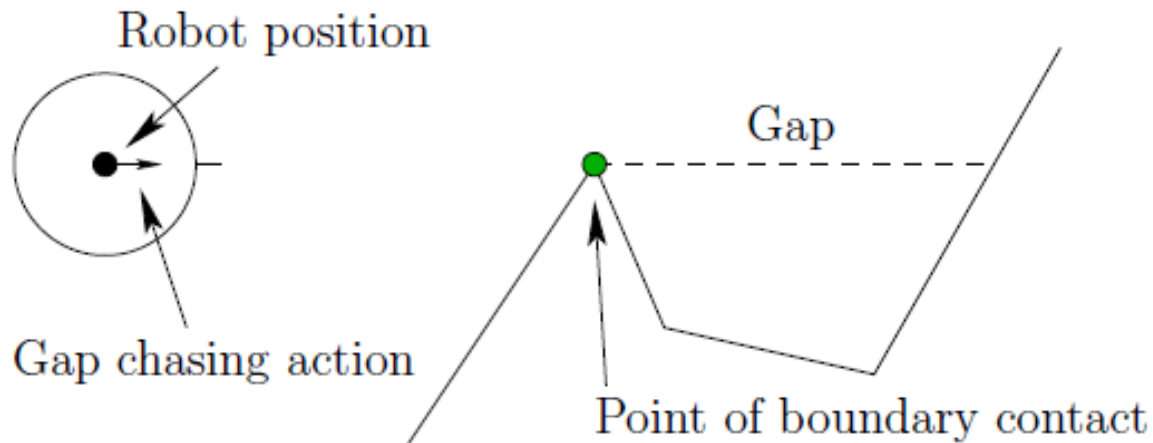


More reading: https://www.cs.cmu.edu/~motionplanning/lecture/Chap2-Bug-Alg_howie.pdf

Gap Navigation Trees (I)

Gap navigation tree (GNT) is another type of navigation algorithm without a map of the environment (similar to bug algorithms)

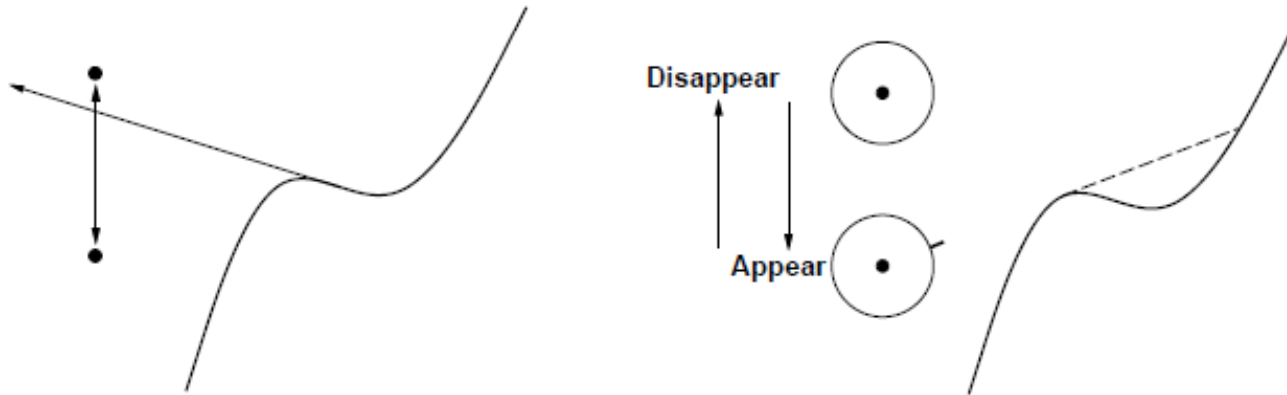
What is a **gap**?



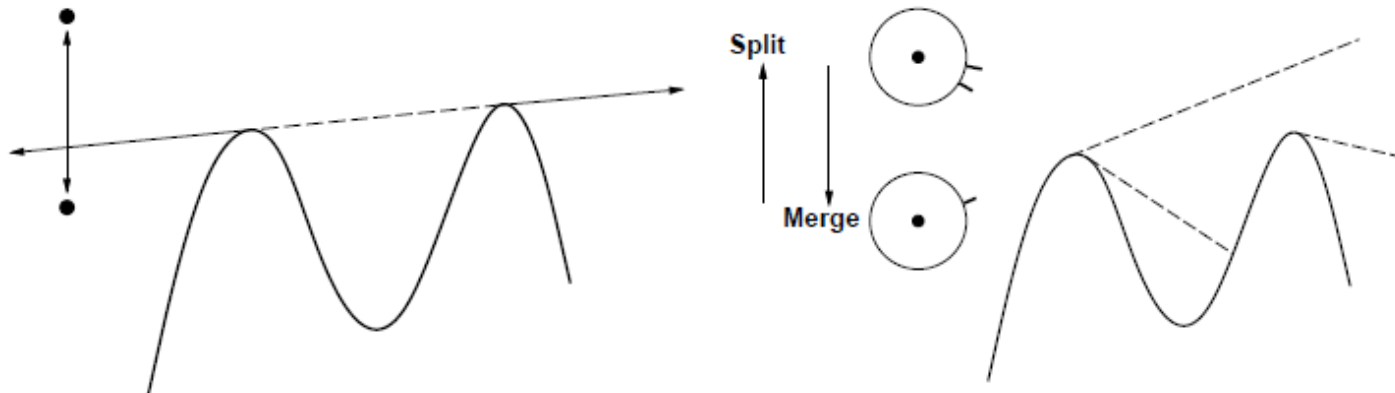
Gap Navigation Trees (II)

Gaps apply to smooth boundary as well. Four types of **critical changes**

⇒ Appear/disappear

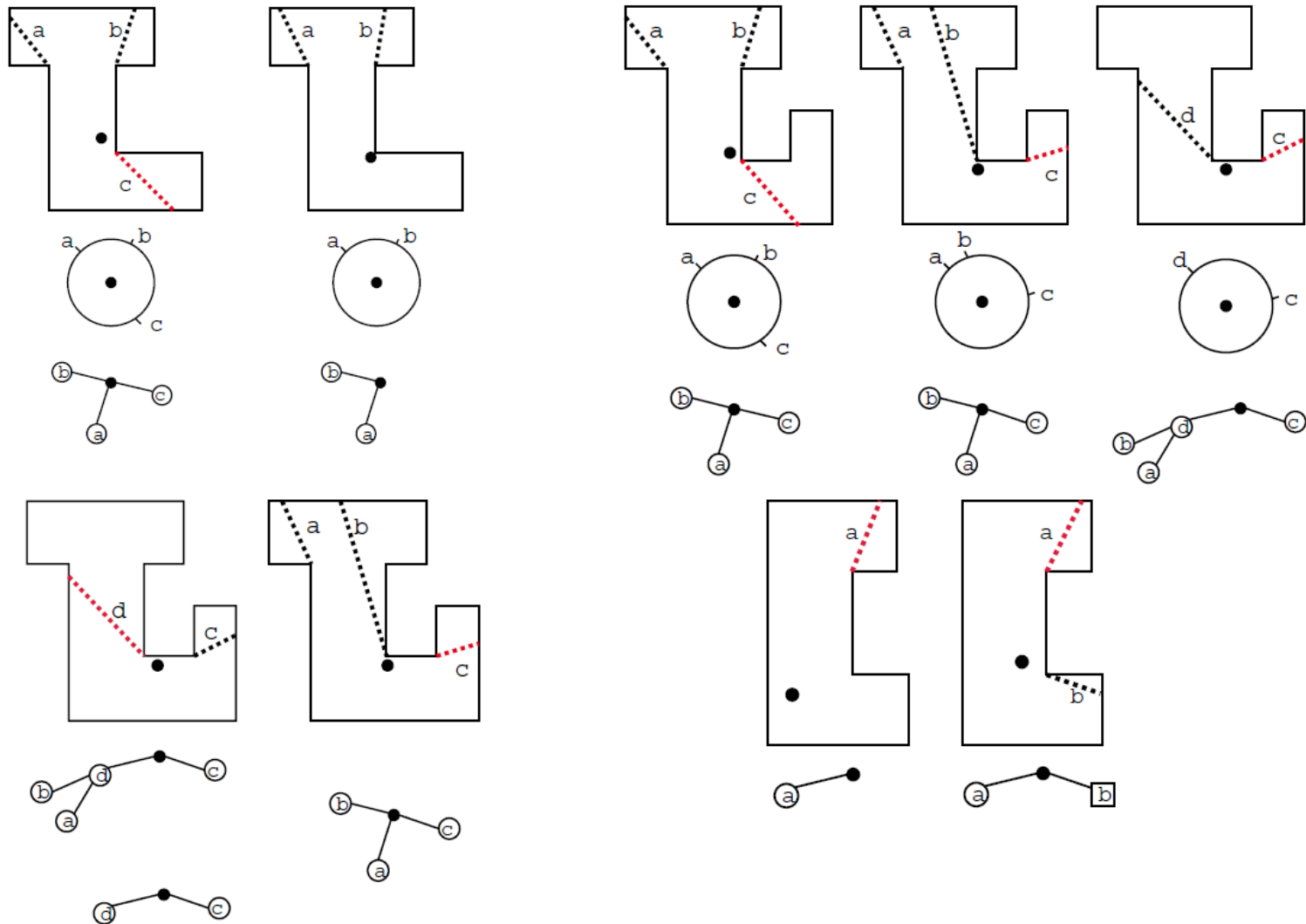


⇒ Split/Merge



Gap Navigation Trees (III)

Building GNT



Gap Navigation Trees (IV)

Using GNT, it becomes possible to navigate regions of an environment without the actual map of the environment via “chasing a gap”

- ⇒ In particular, GNT induces equivalence class on the environments
- ⇒ E.g., all following environments are equivalent under GNT

