# Map Representations, Graphs and Graph Search 

## Outline of the next module on Navigation

- Local planners
- Global localization and planning
- Map representations
- Continuous
- Discrete
- Topological
- Maps as graphs
- Graph Search Algorithms
- Breadth First Search
- Depth First Search
- Dijkstras
- A*

Localization, Sensor and motion models, SLAM
Fast Robots


## Navigation

- Navigation breaks down to: Localization, Map Building, Path Planning



## Map Representation

(a) Building plan
(b) line-based map
(c) occupancy grid-based map
(d) topological map

## Important properties

- Memory allocation
- Computation


## Continuous Representations

- Exact decomposition of the environment
- Used mainly in 2D representations
- Closed-world assumption
- Storage proportional to object density
- Example: Continuous line representations
- Using range finders, we can extract lines/line segments in the environment



## Fixed Decomposition

- Tessellate the world at a fixed resolution
- Approximate features given the resolution
- Most commonly used: Occupancy grid



## Fixed Decomposition



## Adaptive Cell Decomposition

- Adapt cell size to features



## Topological Decomposition

- A topological representation is a graph that specifies nodes and edges
- Nodes denote areas in the environment
- Edges describe environment connectivity
- Robots can...
- ...detect their current position in terms of the nodes of the topological graph
- ...travel between nodes using robot motion



## Topological Decomposition

- A topological representation is a graph that specifies nodes and edges
- Nodes denote areas in the environment
- Edges describe environment connectivity
- Robots can...
- ...detect their current position in terms of the nodes of the topological graph
- ...travel between nodes using robot motion
- Typical for 3D maps

Fast Robots


## How to represent the robot pose?

- Physical robots take up space
- Expand obstacles
- Represent maps in configuration space instead of Euclidean space



## Map Representation Considerations

- The precision of the map must appropriately match the precision with which the robot needs to achieve its goals
- The precision of the map and the type of features represented must match the precision and data types returned by the robot's sensors
- The complexity of the map representation has direct impact on the computational complexity of reasoning about mapping, localization, and navigation


# ECE 4160/5160 MAE 4910/5910 

## Constructing Graphs

## Modelling path planning as a graph search problem

Real world \begin{tabular}{c|c|c|c|c|c|}
Configuration <br>
Space

$\Rightarrow$

Map <br>
Representation

$\quad \Rightarrow$

Graph <br>
Construction

$\quad \Rightarrow$

Graph <br>
Search <br>
\hline
\end{tabular}



Common alternatives

- Optimal control
- Potential fields


## Modelling path planning as a graph search problem



## Graph Construction

- Transform continuous/discrete/topological maps to a discrete graph
- Why?
- Model the path planning problem as a search problem
- Graph theory has lots of tools
- Real-time capable algorithms
- Can accommodate for evolving maps

1. Divide space into simple, connected regions, or "cells"
2. Determine adjacency of open cells
3. Construct a connectivity graph
4. Find cells with initial and goal configuration
5. Search for a path in the connectivity graph to join them
6. From the sequence of cells, compute a path within each cell

- e.g. passing through the midpoints of cell boundaries or by sequence of wall following movements



## Geometry-Based Planners

## Topological Maps

- Good abstract representation
- Tradeoff in \# of nodes
- Complexity vs. accuracy
- Efficient in large, sparse environments
- Loss in geometric precision
- Edges can carry weights
- Con: Limited information



## Fixed Cell Decomposition

(Lab 9-12)


Adaptive Cell Decomposition


## Trapezoidal Cell Decomposition



## Visibility Graphs

- Connect initial and goal locations with all visible vertices
- Connect each obstacle vertex to every visible obstacle vertex
- Remove edges that intersect the interior of an obstacle
- Plan on the resulting graph



## Sampling-Based Planners

- Rather than computing the C-Space explicitly, we sample it
- Often efficient in high dimensional spaces
- Compute if a robot configuration has collisions
- Just requires forward kinematics
- (Local path plans between configurations)
- Examples
- Probabilistic Roadmaps (PRM)
- Rapidly Exploring Random Trees (RRT)


## Probabilistic Roadmaps

- Configurations are sampled by picking coordinates at random



## Probabilistic Roadmaps

- Configurations are sampled by picking coordinates at random
- Sampled configurations are tested for collision



## Probabilistic Roadmaps

- Configurations are sampled by picking coordinates at random
- Sampled configurations are tested for collision
- Each configuration is linked by straight paths to its nearest neighbors



## Probabilistic Roadmaps

- Configurations are sampled by picking coordinates at random
- Sampled configurations are tested for collision
- Each configuration is linked by straight paths to its nearest neighbors
- The collision-free links are retained as local paths to form the PRM



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- Configurations are sampled by picking coordinates at random
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- The start and goal configurations are included as milestones



## Probabilistic Roadmaps

- Configurations are sampled by picking coordinates at random
- Sampled configurations are tested for collision
- Each configuration is linked by straight paths to its nearest neighbors
- The collision-free links are retained as local paths to form the PRM
- The start and goal configurations are included as milestones
- The PRM is searched for a path from start to goal



## Probabilistic Roadmaps

- Considerations
- Single query/multi query
- How are nodes placed?
- Uniform sampling strategies


## realtime robotics

- Non-uniform sampling strategies
- How are local neighbors found?
- How is collision detection performed?
- Dominates time consumption in PRMs



## Rapidly Exploring Random Trees (RRT)



1. Maintain a tree rooted at the starting point
2. Choose a point at random from free space
3. Find the closest configuration already in the tree
4. Extend the tree in the direction of the new configuration

## Rapidly Exploring Random Trees (RRT) - Uniform/biased sampling



Aaron Becker, UH, Wolfram Player example

## Rapidly Exploring Random Trees (RRT) - Considerations

- Sensitive to step-size ( $\Delta \mathrm{q}$ )
- Small: many nodes, closely spaced, slowing down nearest neighbor computation
- Large: Increased risk of suboptimal plans / not finding a solution
- How are samples chosen?
- Uniform sampling may need too many samples to find the goal
- Biased sampling towards goal can ease this problem
- How are closest neighbors found?
- How are local paths generated?
- Variations
- RRT Connect, A*-RRT, Informed RRT*, Real-Time RRT*, Theta*-RRT, etc.


# ECE 4160/5160 <br> MAE 4910/5910 

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## Fast Robots Graph Search

## Modelling path planning as a graph search problem


https://pythonrobotics.readthedocs.io/en/latest/modules /path planning.html\#basic-rrt

- Breadth first
- Depth first
- Dijstra
- A*


## Graph Search

- What is the simplest thing to do?
- Random or brute force search
- Other methods?
- Uninformed search
- Depth First Search (DFS)
- Breadth First Search (BFS)
- Dijsktra's Search (LCFS)
- Informed Search
- Greedy
- A*
- (and many more)



## Comparing Search Algorithms

## Vocabulary

- Node, edge, parents/children, branching factor, depth


## Definitions

- Complete
- Guaranteed to find a solution in finite time
- Time complexity
- Worst-case run time
- Space complexity
- Worst-case memory

- Optimality
- A search is optimal if it is complete, and only returns cost-minimizing solutions


## Algorithms and Search

Search order: N, E, S, W
Find a goal

- What is the simplest thing to do?
- Random or brute force search
- How many grid traversals will brute force take?
- First establish a search order
- Advance x first, then increment $y$ and decrease $x$, etc.



## Algorithms and Search

Search order: N, E, S, W

- What is the simplest thing to do?
- Random or brute force search
- Other methods?
- Depth First Search (DFS)

Find a goal


## Algorithms and Search

Search order: N, E, S, W
Find a goal

- What is the simplest thing to do?
- Random or brute force search
- Other methods?
- Depth First Search (DFS)
- Breadth First Search (BFS)

| 10 | 14 |  |  |
| :---: | :---: | :---: | :---: |
| 6 | 11 | $\hat{\beta}$ |  |
| 3 | 7 | 12 |  |
| 1 | 4 | 8 | 13 |
| $S$ | 2 | 5 | 9 |
| $X$ |  |  |  |

## Depth First Search (DFS)

Search order: N, E, S, W
Find a goal


## Depth First Search (DFS)

Search order: N, E, S, W
Find a goal


## Breadth First Search (BFS)

Search order: N, E, S, W
Find a goal


## Search Algorithms, General

Find a goal

- Common graph structure
- For every node, n
- you have a set of actions, a
- that moves you to a new node, n'



## Search Algorithms, General

```
n = state(init)
frontier.append(n)
while(frontier not empty)
    n = pull state from frontier
    append n to visited
    if n = goal, return solution
    for all actions in n
        n' = a(n)
        if n' not visited
        append n' to frontier
```



How much space to allocate to your buffers?

## Depth First Search (DFS)

```
frontier visited
```

```
n = state(init)
```

n = state(init)
frontier.append(n)
frontier.append(n)
while(frontier not empty)
while(frontier not empty)
n = pull state from frontier
n = pull state from frontier
append n to visited
append n to visited
if n = goal, return solution
if n = goal, return solution
for all actions in n
for all actions in n
n'}=\textrm{a}(\textrm{n}
n'}=\textrm{a}(\textrm{n}
if n' not visited
if n' not visited
append n' to frontier

```
            append n' to frontier
```

frontier



## Depth First Search (DFS)

```
n = state(init)
frontier.append (n)
while(frontier not empty)
    n = pull state from frontier
    append n to visited
    if n = goal, return solution
    for all actions in n
        n' = a(n)
        if n' not visited
            append n' to frontier
```


frontier
visited


## Depth First Search (DFS)

```
n = state(init)
frontier.append(n)
while(frontier not empty)
    n = pull state from frontier
    append n to visited
    if n = goal, return solution
    for all actions in n
        n' = a(n)
        if n' not visited
            append n' to frontier
```


visited


## Depth First Search (DFS)

```
        n = state(init)
frontier.append(n)
while(frontier not empty)
    n = pull state from frontier
    append n to visited
    if n = goal, return solution
    for all actions in n
        n' = a(n)
        if n' not visited
            append n' to frontier
```

| 0,0 |
| :---: |
| 0,1 |
| 0,2 |

        frontier visited
    

## Depth First Search (DFS)

```
n = state(init)
frontier.append(n)
while(frontier not empty)
    n = pull state from frontier
    append n to visited
    if n = goal, return solution
    for all actions in n
        n' = a(n)
        if n' not visited
            append n' to frontier
```


1,2
frontier visited


## Depth First Search (DFS)

```
n = state(init)
frontier.append(n)
while(frontier not empty)
    n = pull state from frontier
    append n to visited
    if n = goal, return solution
    for all actions in n
        n' = a(n)
        if n' not visited
        append n' to frontier
```



## Frontier Buffer?

- Last-In First-Out (LIFO) Buffer


## Depth First Search (DFS)

- Is it complete?
- Yes, but only on finite graphs
- What is the time complexity?
- $\mathrm{O}\left(b^{m}\right)$
- What is the space complexity?
- $\mathrm{O}(b m)$



## Breadth First Search (BFS)

```
n = state(init)
frontier.append(n)
while(frontier not empty)
    n = pull state from frontier
    if n is goal, return solution
    for all actions in n
        n' = a(n)
        if n' not visited
            append n' to frontier for all actions in \(n\) \(\mathrm{n}^{\prime}=\mathrm{a}(\mathrm{n})\)
if \(n^{\prime}\) not visited
append \(n^{\prime}\) to frontier
```

frontier
0,0
0,1

1,0
0,2
1,1
2,0
0,3
visited

| 0,0 |
| :---: |
| 0,1 |
| 1,0 |

0,2


## Breadth First Search (BFS)

- Is it complete?
- Yes, as long as $b$ is finite
- Is it optimal?
- Yes
- What is the time complexity?
- $\mathrm{O}\left(b^{m}\right)$
- What is the space complexity?
- $\mathrm{O}\left(b^{m}\right)$


```
frontier
visited
```

0,0
0,1

1,0
0,2



Type of Buffer?

## BFS: Memory and Computation

Frontier size:

- 4
- 12
- 36




## Uninformed Search Algorithms, General

- When is DFS appropriate?
- If the memory is restricted
- If solutions tend to occur at the same depth in the tree
- When is DFS inappropriate?
- If some paths have infinite length / if the graph contains cycles
- If some solutions are very deep, while others are very shallow
- When is BFS appropriate?
- If you need to find the shortest path
- If memory is not a problem
- If some solutions are shallow
- If there might be infinite paths
- When is BFS inappropriate?
- If memory is limited / if the branching factor is very large

If solutions tend to be located deep in the tree

## ECE4160/5160 - MAE 4910/5910 Fast Robots

- Is BFS / DFS possible for your task on the Artemis?
- What is the maximum branching factor?
- $b=4$
- What is the longest path?
- $m \sim 20^{*} 20=400$
- Depth First Search
- Frontier: $\mathrm{O}(b m)=1,600$ nodes
- Float -> 6.4 kB
- Artemis memory?
- 1MB flash and 384k RAM
- Breadth First Search

- Frontier: $\mathrm{O}\left(b^{m}\right)=4^{20^{*} 20}=6.7 \mathrm{e} 240$ nodes


## BFS: Memory and Computation

Frontier size:

- 4
- 12
- 36





## Lowest-Cost First Search

- Consider parent cost!


What node to expand next?

Data structure

- n.state
- n.parent
- n.cost
- n.action

What cost heuristic could we add?

- Go straight, cost 1
- Turn one quadrant, cost 1

|  | $(1,4)$ | $(2,4)$ | $(3,4)$ |
| :---: | :---: | :---: | :---: |
|  | $(1,3)$ | R | $(3,3)$ |
|  | $(1,2)$ | $(2,2)$ | $(3,2)$ |
|  | $G$ | $(2,1)$ | $(3,1)$ |
|  |  | $(2,0)$ |  |

## Lowest-Cost First Search

- Consider parent cost!

```
```

n = state(init)

```
```

n = state(init)
frontier.append(n)
frontier.append(n)
while(frontier not empty)
while(frontier not empty)
n = pull state from frontier
n = pull state from frontier
visited.append(n)
visited.append(n)
if n = goal, return solution
if n = goal, return solution
for all actions in n
for all actions in n
n' = a(n)
n' = a(n)
if n' not visited
if n' not visited
priority = heuristic(goal,n')
priority = heuristic(goal,n')
frontier.append(priority)

```
```

            frontier.append(priority)
    ```
```


## we add?

- Go straight, cost 1
- Turn one quadrant, cost 1

|  | $(1,4)$ | $(2,4)$ | $(3,4)$ |
| :---: | :---: | :---: | :---: |
|  | $(1,3)$ | R | $(3,3)$ |
|  | $(1,2)$ | $(2,2)$ | $(3,2)$ |
|  | $G$ | $(2,1)$ | $(3,1)$ |
|  |  | $(2,0)$ |  |

## Lowest-Cost First Search

- Is it complete?
- Yes, as long as path costs are positive
- What is the time complexity?
- $\mathrm{O}\left(b^{m}\right)$
-What is the space complexity?
- $\mathrm{O}\left(b^{m}\right)$



## Could we be smarter?

- Sure, you know the graph and you know the goal is!
- ...Informed search
- Consider parent cost, and..
- ..estimate the shortest path to the "goal"
- Assign a value to the frontier
- Pick frontier closest to the goal (minimize distance)


## Informed Search

Search order: N, E, S, W

- Greedy Search



## Informed Search

- Greedy Search

```
n = state(init)
frontier.append(n)
while(frontier not empty)
    n = pull state from frontier
    visited.append(n)
    if n = goal, return solution
    for all actions in n
        n' = a(n)
        if n' not visited
            priority = heuristic(goal,n')
            frontier.append(priority)
```



## Informed Search

Search order: N, E, S, W

- Greedy Search
- Complete?
- No
- Time complexity?
- $\mathrm{O}\left(b^{m}\right)$
- Space complexity?
- $\mathrm{O}\left(b^{m}\right)$
- Optimal?
- no...



## Search Algorithms, General

- Breadth First Search
- Complete and optimal
- ...but searches everything
- Lowest-Cost First Algorithm Considers parent cost
- Complete and optimal
- ...but it wastes time exploring in directions that aren't promising
- Greedy Search Considers goal
- Complete (in most cases)
- ...only explores promising directions


## Informed Search

Search order: N, E, S, W
Find a treasure

- A* ("A-star")

$$
\begin{aligned}
& \mathrm{n}=\text { state (init) } \\
& \text { frontier.append ( } \mathrm{n} \text { ) }
\end{aligned}
$$

while(frontier not empty)

$$
\mathrm{n}=\text { pull state from frontier }
$$

$$
\text { if } \mathrm{n}=\text { goal, return solution }
$$

$$
\text { for all actions in } n
$$

$$
n^{\prime}=a(n)
$$

$$
\text { if ( ( } n^{\prime} \text { not visited or }
$$

$$
\text { (visited and } n^{\prime} \text {.cost } \text {, n_old. cost) ) }
$$ priority $=$ heuristic (goal, $\mathrm{n}^{\prime}$ ) +cost frontier.append(prioricy) visited.append ( $\mathrm{n}^{\prime}$ )

## Informed Search

Search order: N, E, S, W
Find a goal

- A* ("A-star")
- Cost and goal heuristic



## A* Search

- What if the heuristic is too optimistic?
- Estimated cost < true cost
- What if the heuristic is too pessimistic?
- Estimated cost > true cost
- No longer guaranteed to be optimal
- What if the heuristic is just right?
- Pre-compute the cost between all nodes
- Feasible for you?



## Informed Search

- $A^{*}$ ("A-star")
- Cost and goal heuristic

- Complete?
- Yes!
- Time Complexity
- $O\left(b^{m}\right)$
- Space Complexity
- $\mathrm{O}\left(b^{m}\right)$

Optimal?

- Yes, if the heuristic is admissible!


## Summary

| LCFS | minimum path |  |  |
| :---: | :---: | :---: | :---: |
| 7 | 12 | 15 |  |
| 4 | 10 | 2 |  |
| 2 | 8 | 13 |  |
| 1 | 5 | 11 | 14 |
| 5 | 3 | 6 | 9 |



A*

| 10 |  |  |  |
| :---: | :---: | :---: | :---: |
| 6 | 9 | 7 |  |
| 3 | 7 | 11 |  |
| 1 | 4 | 8 |  |
| 5 | 2 | 5 |  |




