

ENPM661: Planning for Autonomous Robots

Splines Revisited, Graphs, Queues, Graph Search

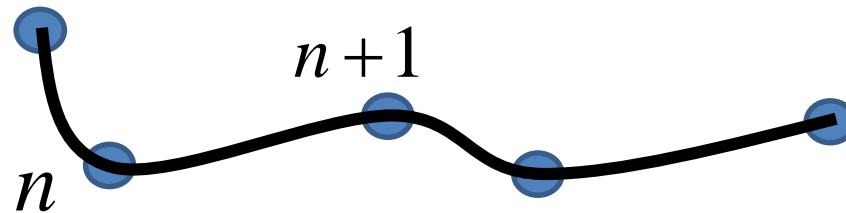
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Cubic Hermite splines, revisited

- Multi-segment curve $g(x)$
- Each segment is a 3rd degree polynomial
 - At endpoints of the n -th segment we choose:
 - Locations: p_n and p_{n+1}
 - Tangent vectors: d_n and d_{n+1}



Cubic Hermite splines, revisited

- Focus on a single segment
- In general we can use a parametric form

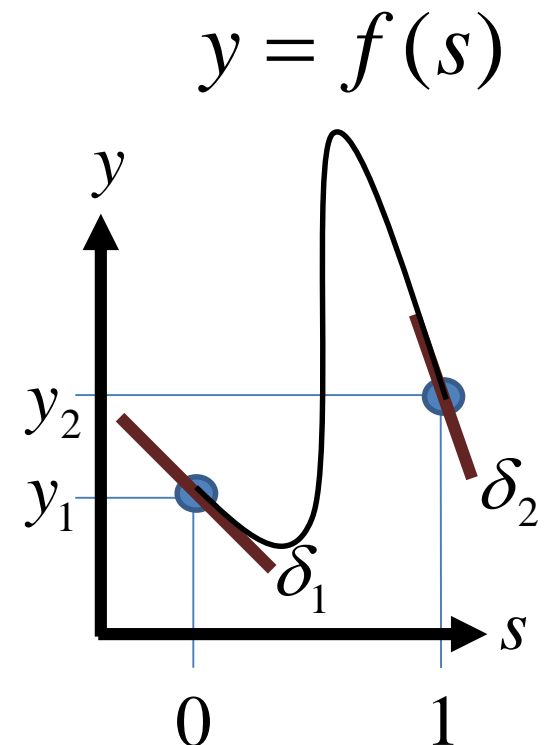
$$(x, y, z, \dots) = f(s)$$

- Here we will just focus on the 1-D case

$$y = f(s)$$

- We assume s goes from 0 to 1
- We get to pick $y_1, y_2, \delta_1, \delta_2$

$$\delta_1 = \frac{dy_1}{ds} \quad \delta_2 = \frac{dy_2}{ds}$$



Cubic Hermite splines, revisited

- Want to find $y = f(s) = as^3 + bs^2 + cs + d$ that goes through y_1 and y_2 at δ_1 and δ_2 .
- Do this by solving:

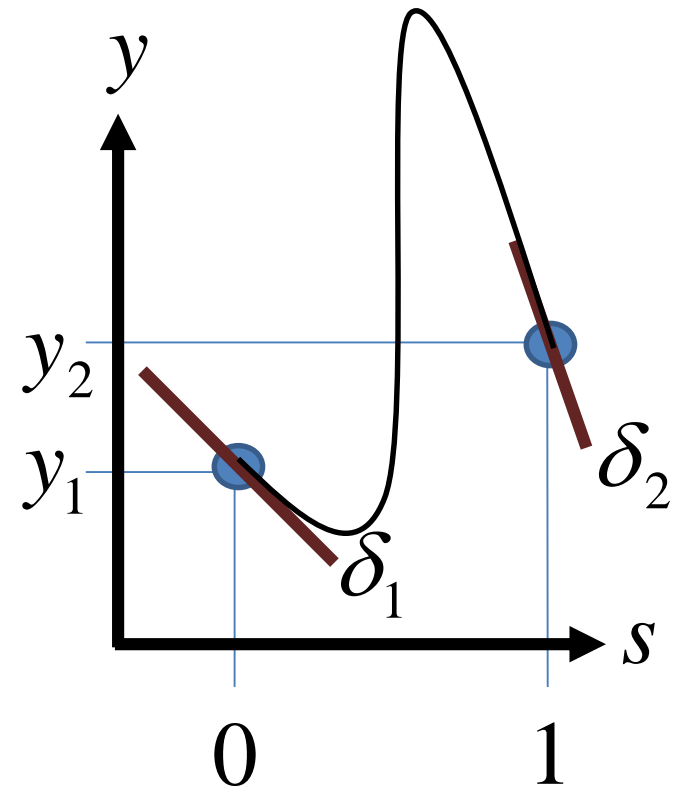
$$f(0) = y_1 = d$$

$$f(1) = y_2 = a + b + c + d$$

$$f'(0) = \delta_1 = c$$

$$f'(1) = \delta_2 = 3a + 2b + c$$

$$y = f(s) = as^3 + bs^2 + cs + d$$



Cubic Hermite splines, revisited

- Want to find $y = f(s) = as^3 + bs^2 + cs + d$ that goes through y_1 and y_2 at δ_1 and δ_2 .

- Do this by solving:

$$f(0) = y_1 = d$$

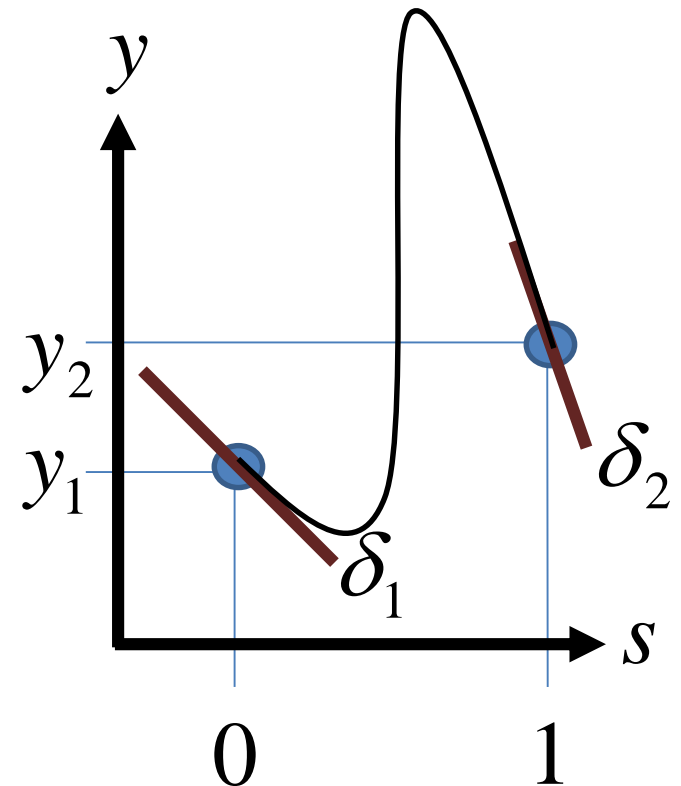
$$f(1) = y_2 = a + b + c + d$$

$$f'(0) = \delta_1 = c$$

$$f'(1) = \delta_2 = 3a + 2b + c$$

$$y = f(s) = as^3 + bs^2 + cs + d$$

$$H = \begin{bmatrix} y_1 \\ y_2 \\ \delta_1 \\ \delta_2 \end{bmatrix} \quad M = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{bmatrix} \quad C = \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix}$$



- Solve: $H=MC$ for C

Cubic Hermite splines, revisited

- Want to find $y = f(s) = as^3 + bs^2 + cs + d$

$$M^{-1}M = \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} * \begin{bmatrix} 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

- Solve: $H = MC$ for C

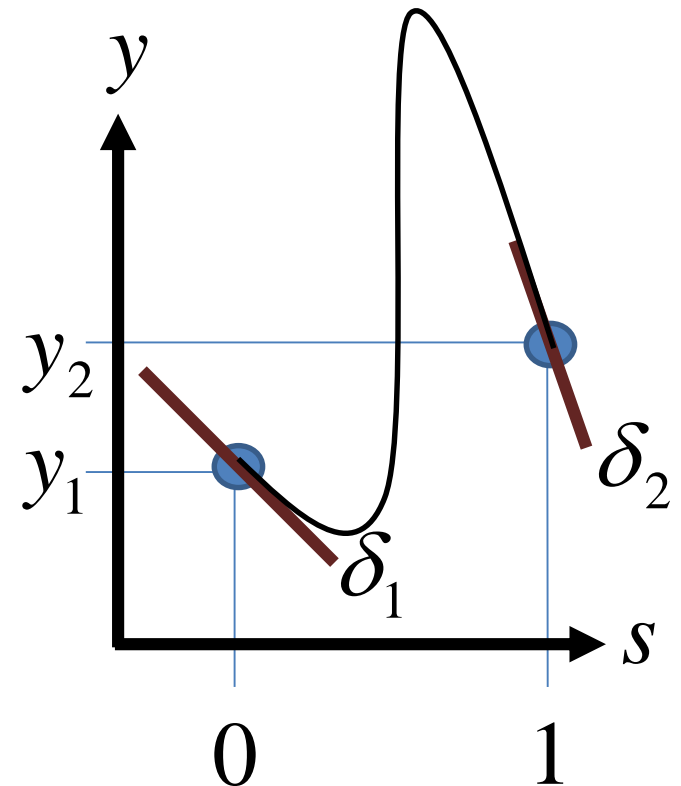
$$H = MC$$

$$M^{-1}H = M^{-1}MC$$

$$M^{-1}H = C$$

$$C = M^{-1}H$$

$$\begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} * \begin{bmatrix} y_1 \\ y_2 \\ \delta_1 \\ \delta_2 \end{bmatrix}$$



Switch to code

Cubic Hermite splines, revisited

- If all we care about is a single value $y = f(s)$

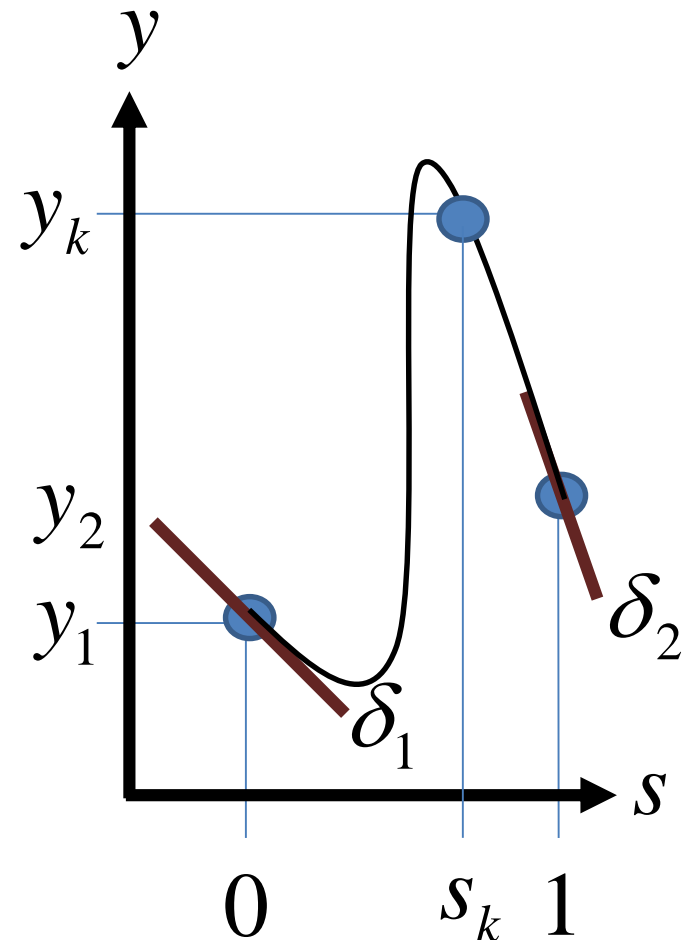
$$S = \begin{bmatrix} s^3 & s^2 & s & 1 \end{bmatrix}$$

$$f(y) = SC$$

$$f(y) = SM^{-1}H$$

$$f(y) = \begin{bmatrix} s^3 & s^2 & s & 1 \end{bmatrix} * \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} * \begin{bmatrix} y_1 \\ y_2 \\ \delta_1 \\ \delta_2 \end{bmatrix}$$

- Inefficient for more than 1 point
- Yes, S is a row vector
- (notation change vs. last class)



Switch to code

Cubic Hermite splines, revisited

■ Basis function interpretation

$$f(s) = SC$$

$$f(s) = C^T S^T$$

$$f(s) = C^T \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} S^T$$

$$f(s) = C^T M^T (M^{-1})^T S^T$$

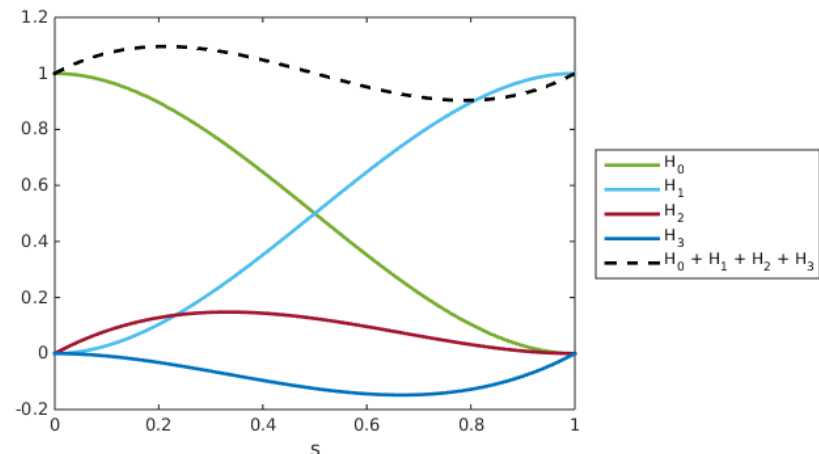
$$f(s) = (C^T M^T) ((M^{-1})^T S^T)$$

$$f(s) = h \mathbf{H}_{(s)}$$

$$h = \begin{bmatrix} h_1 \\ h_2 \\ h_3 \\ h_4 \end{bmatrix} = C^T M^T = [a \quad b \quad c \quad d] * \begin{bmatrix} 0 & 1 & 0 & 3 \\ 0 & 1 & 0 & 2 \\ 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 \end{bmatrix}$$

$$\mathbf{H}_{(s)} = \begin{bmatrix} H_1(s) \\ H_2(s) \\ H_3(s) \\ H_4(s) \end{bmatrix} = (M^{-1})^T S^T = \begin{bmatrix} 2 & -3 & 0 & 1 \\ -2 & 3 & 0 & 0 \\ 1 & -2 & 1 & 0 \\ 1 & -1 & 0 & 0 \end{bmatrix} * \begin{bmatrix} s^3 \\ s^2 \\ s \\ 1 \end{bmatrix}$$

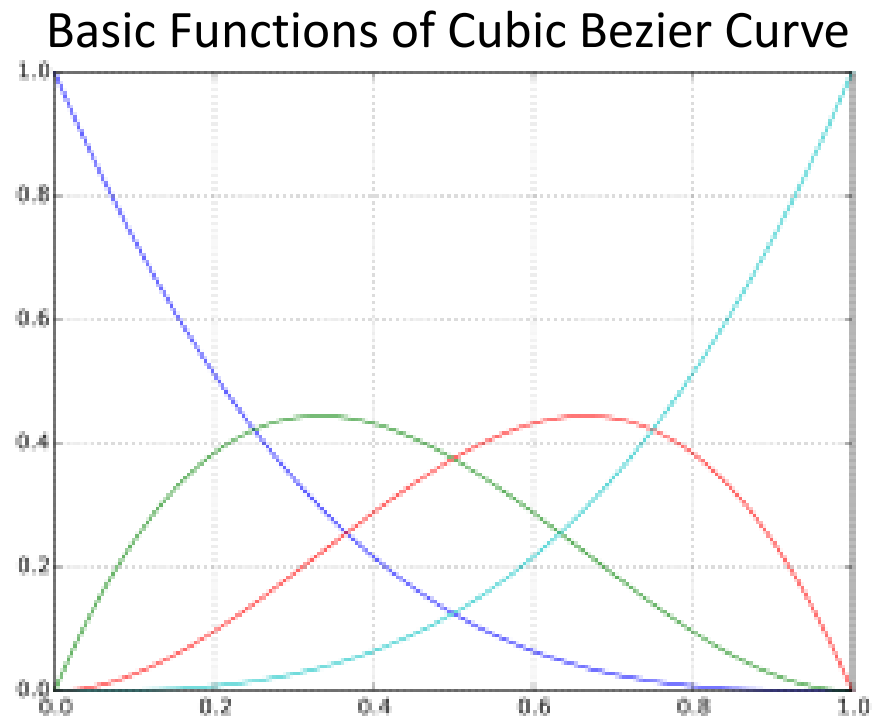
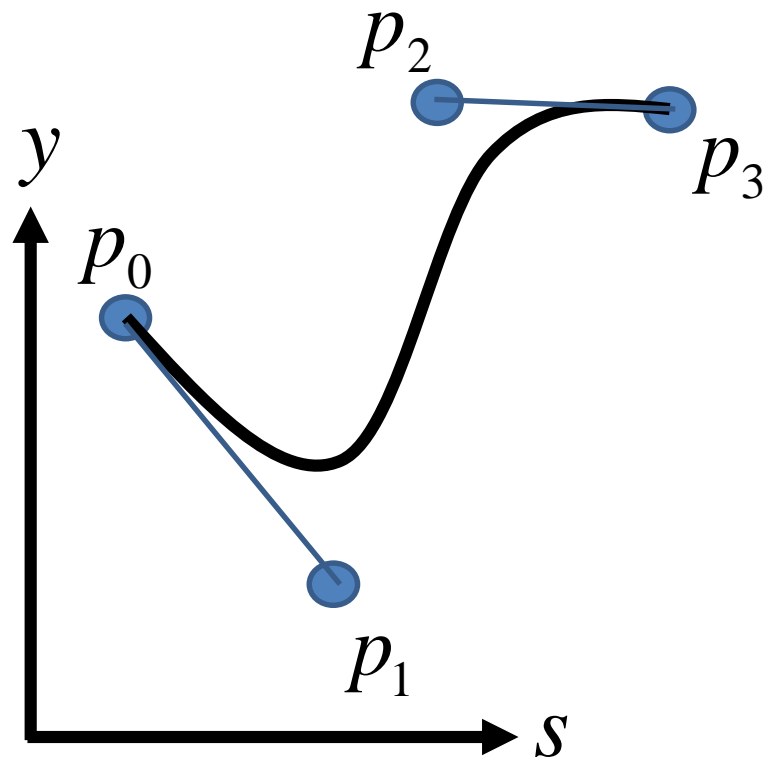
- $\mathbf{H}_{(s)}$ is given and we choose the weights h by picking $y_1, y_2, \delta_1, \delta_2$



Switch to code

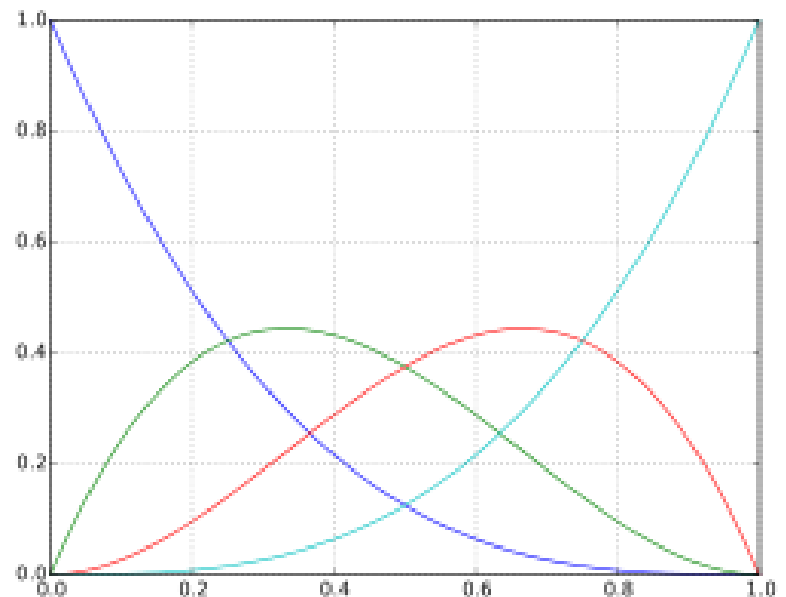
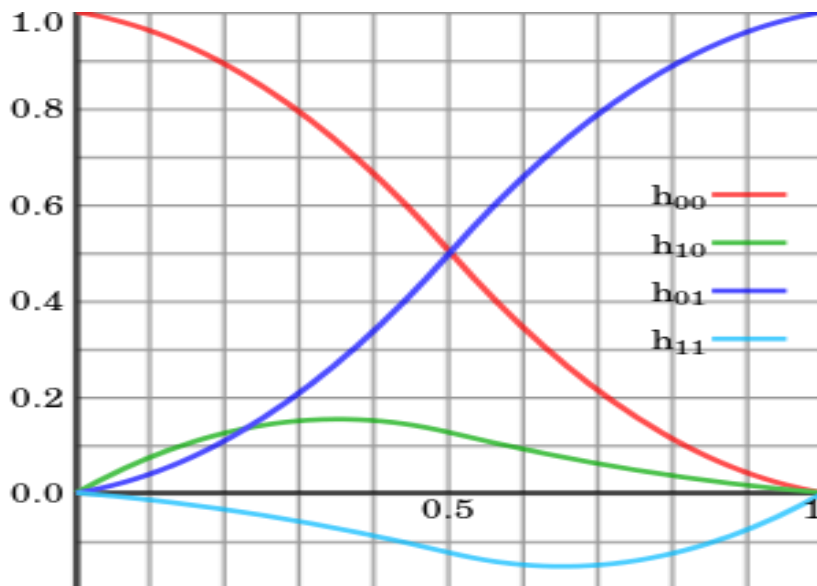
Bezier Curves, revisited

- Bezier: Basis function interpretation
- Bezier Curves are similar, but we use a different set of basis functions.
 - We pick y_0, y_1, y_2, y_3 which determines the weights h on a different set of basis functions $\mathbf{H}_{(s)}$.



B-Splines Curves, revisited

- B-Splines: Basis function interpretation
- B-splines divide a curve into segments, such that the resulting spline has continuity to order n .
 - We choose n and {points, tangents, knots, etc.}.
 - There is always a way to interpret what is going on (for each segment) as choosing $n+1$ weights on a set of basis functions.



Assignments

Homework Assignment #H1

- Online on Campus.
- Due: February 20, 11:59PM (one week)
 - Upload to Campus
- Each student does their own assignment.
 - I don't care if you talk to each other about the homework...
 - but everyone must do the assignment on their own.
- 5 Questions:
 1. Hermite Spline
 2. Dubins Curve
 3. Coordinate Frame Transforms
 4. Queues (covered shortly)
 5. Graph Search (covered shortly)

Project

- Progress?
- Who still is trying to figure out a topic?

On to new things....

Reading that Corresponds to today's lecture



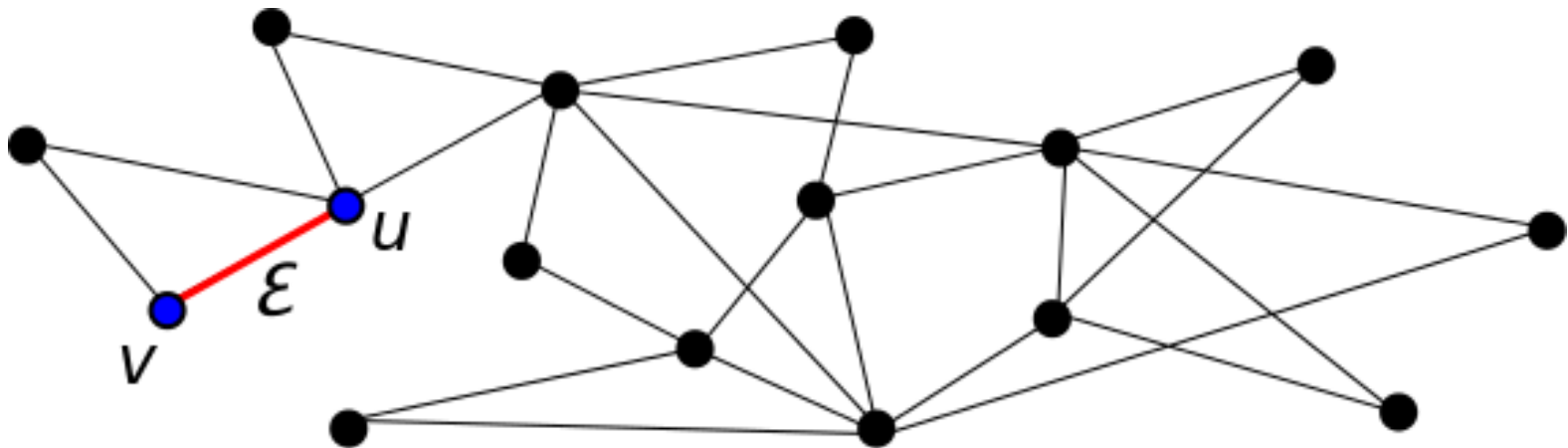
(chapters based on website version)

Chapter 2.0-2.2, Discrete Planning (first half)

Graphs

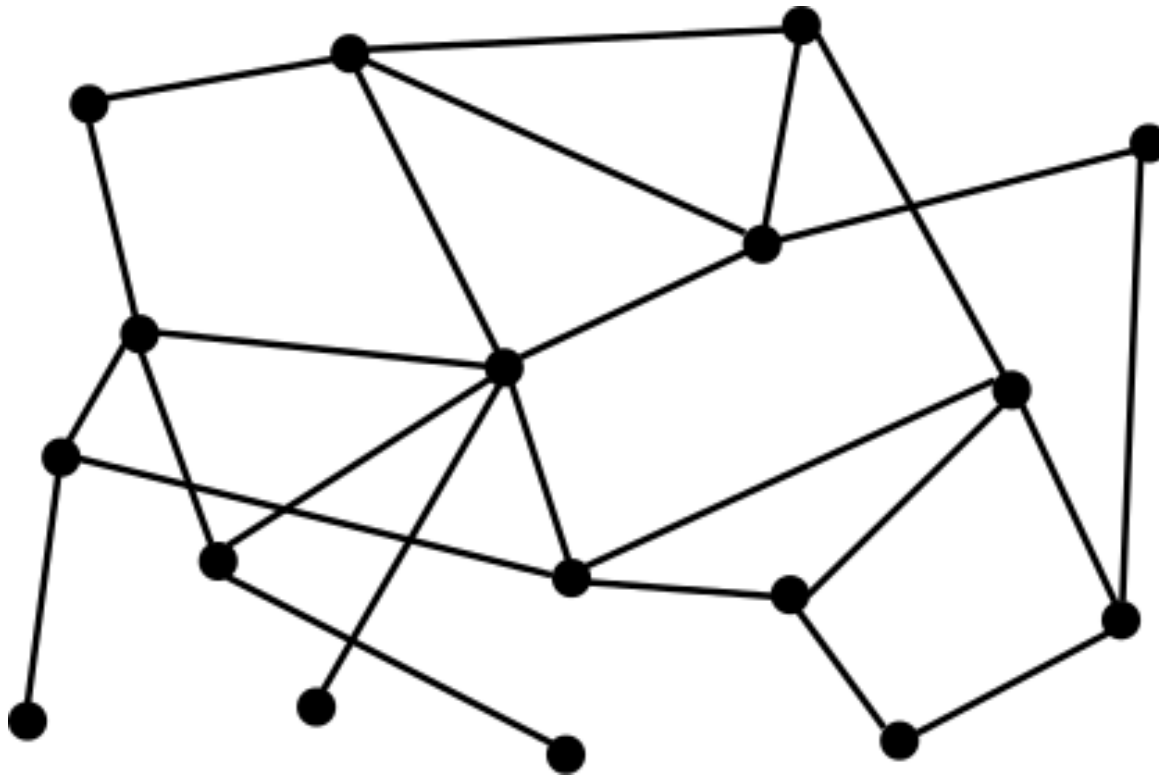
Graphs

- Node v
- Node set $V = \bigcup_i \{v_i\}$
- Edge $\varepsilon = (v, u)$ where $v, u \in V$
- Edge set $E = \bigcup_k \{\varepsilon_k\}$
- Graph $G = (V, E)$



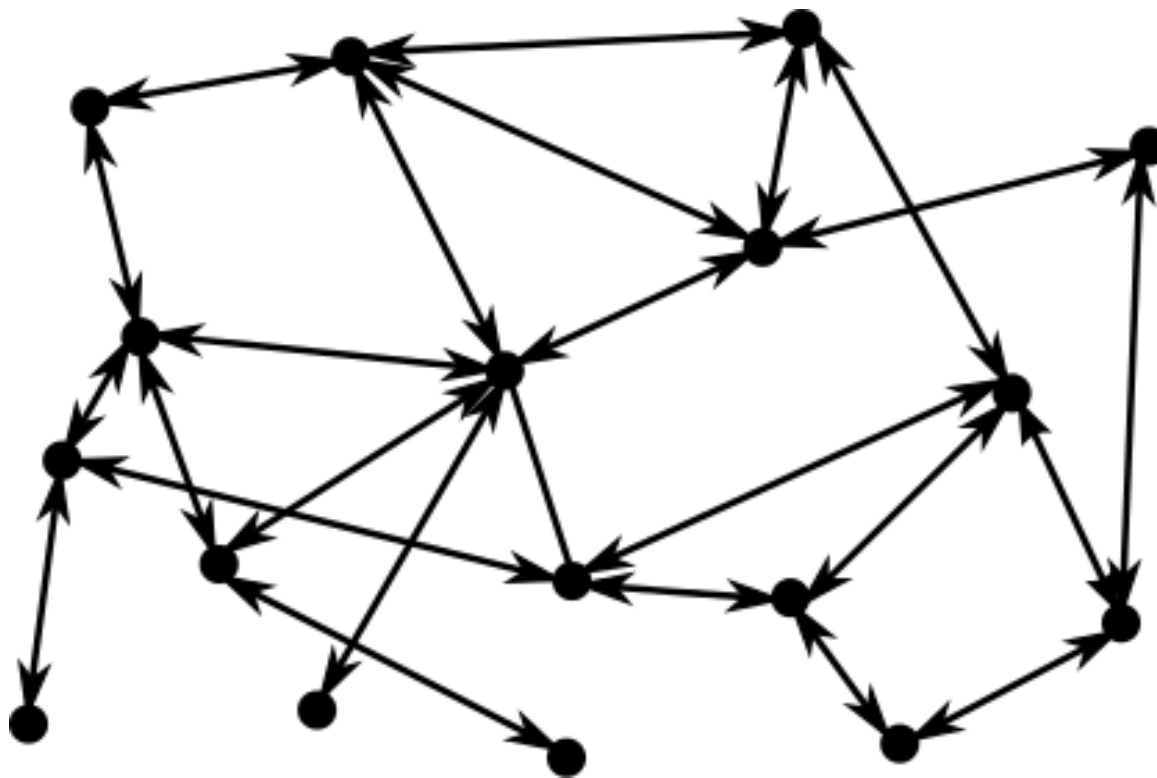
Undirected Graphs

- It is possible to go either way along an edge
- Or, alternatively, $(v, u) \in E \Rightarrow (u, v) \in E$



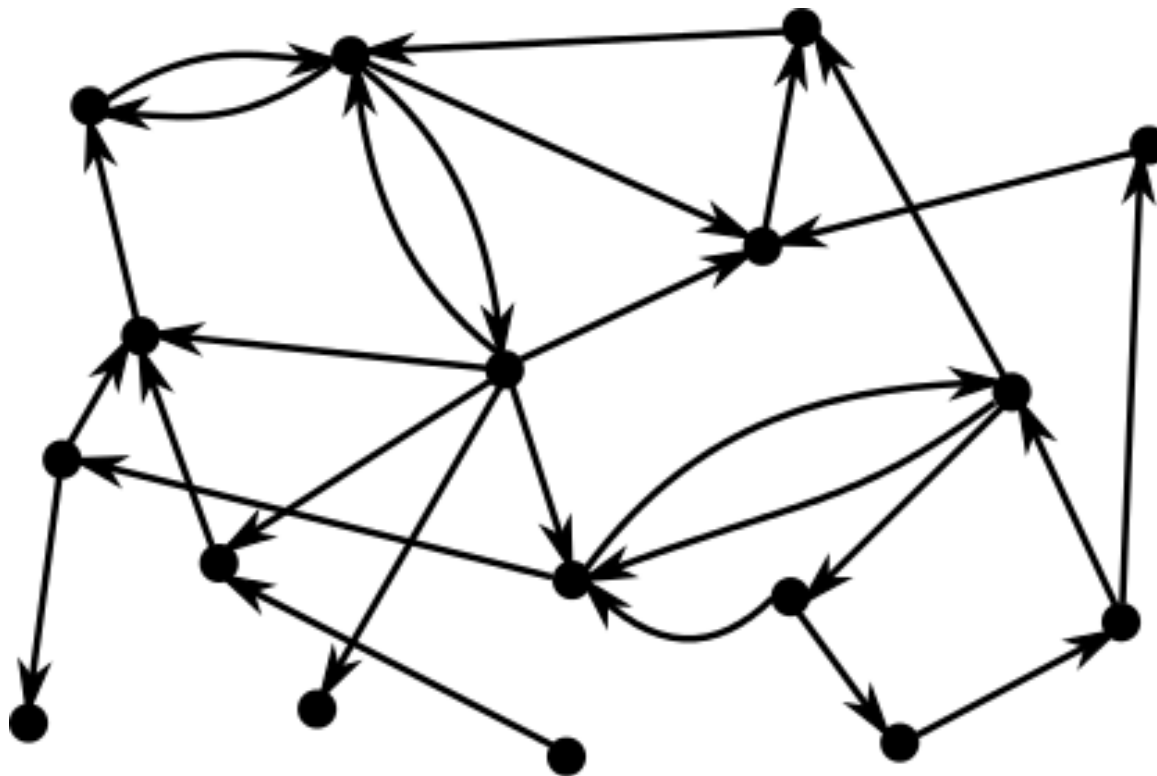
Undirected Graphs

- It is possible to go either way along an edge
- Or, alternatively, $(v, u) \in E \Rightarrow (u, v) \in E$



Directed Graphs

- Edges only go one way
- Or, alternatively, $(v, u) \in E \not\Rightarrow (u, v) \in E$

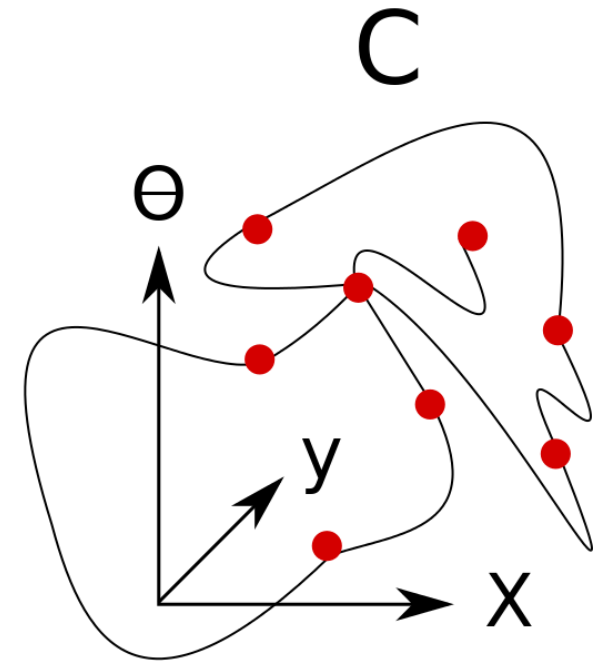
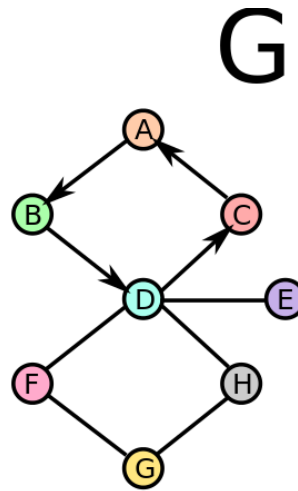
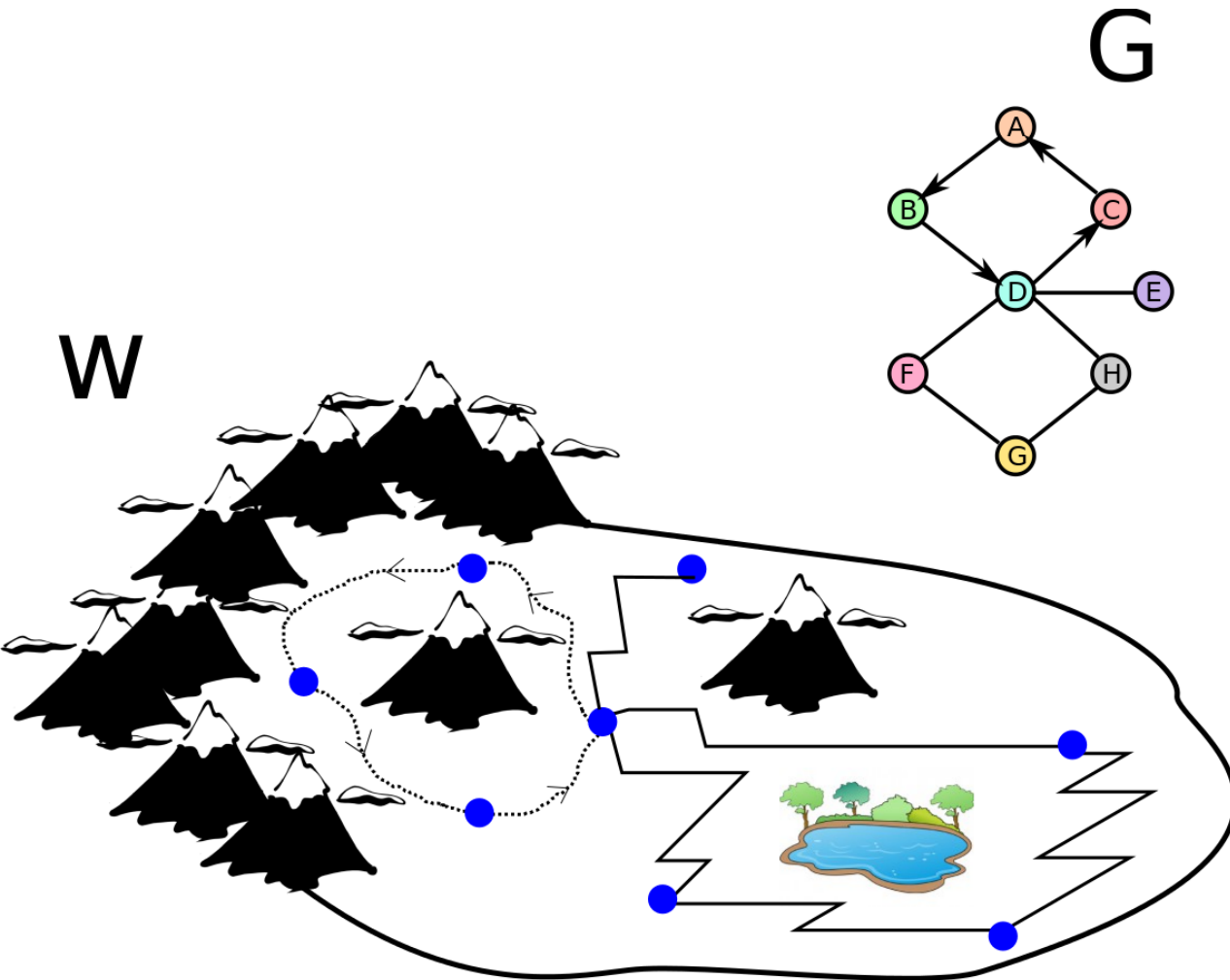


My soapbox about using graphs for motion planning

- Graphs are a theoretical tool.
- Edges represent trajectories (or control policies) that the robot can follow.
- Edges are NOT the same thing as trajectories.
 - **Edge** are a **theoretical construct** for reasoning about whether or not we can get from u to v .
 - **Trajectories** are specific **geometric curves** through the configuration space from one configuration X to another Y .
- Nodes are NOT configurations.
 - Nodes u, v **represent** configurations X, Y in theory and code.

My soapbox about using graphs for motion planning

- Graphs are a theoretical tool.
- Edges *represent* trajectories.
- Nodes *represent* configurations.



My soapbox about using graphs for motion planning

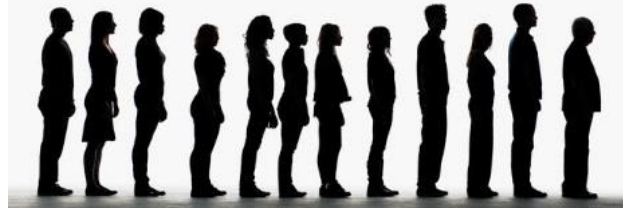
- That said...
- We often draw trajectories and talk about edges...
 - Edges are purely theoretical we might as well look at the trajectories they represent to build intuition.
- In fact, we often project trajectories to the workspace for visualization (even though trajectories exist in the configuration space).
- Always remember:
 - Planning happens in the configuration space.
 - Edges are not trajectories (but represent them).

Data Structures

Data Structures

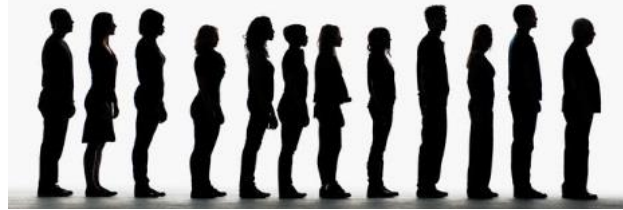
- Low-level
 - int, float, double, char, etc.
- Mid-level
 - array, vector, matrix
 - linked-list, double linked list.
- High level
 - Queue, Stack, Heap
 - Tree, Graph

Queue



- Technically a “First-In-First-Out Queue” or FIFO-Queue
- Q Queue
- Q.INSERT(v) adds node v to the “back”
 - Q.PUSH_BACK()
- Q.TOP() returns the node at the “front”
 - Q.FRONT()
- Q.POP() returns the node at the “front” and removes it from Q
 - Q.POP_FRONT()

Queue



- Technically a “First-In-First-Out Queue” or FIFO-Queue

- What happens:

Q.INSERT(A)
Q.INSERT(B)
Q.TOP()
Q.INSERT(C)
Q.POP()
Q.INSERT(D)
Q.POP()
Q.TOP()
Q.POP()
Q.INSERT(E)

Q.INSERT(A)
Q.POP()
Q.POP()

Stack



- Technically a “Last-In-First-Out Queue” or LIFO-Queue
- Q
- Q.INSERT(v)
 - Q.PUSH_FRONT()
 adds node v to the “top”
- Q.TOP()
 - Q.FRONT()
 returns the node at “top”
- Q.POP()
 - Q.POP_FRONT()
 returns the node at “top”
and removes it from Q

Stack

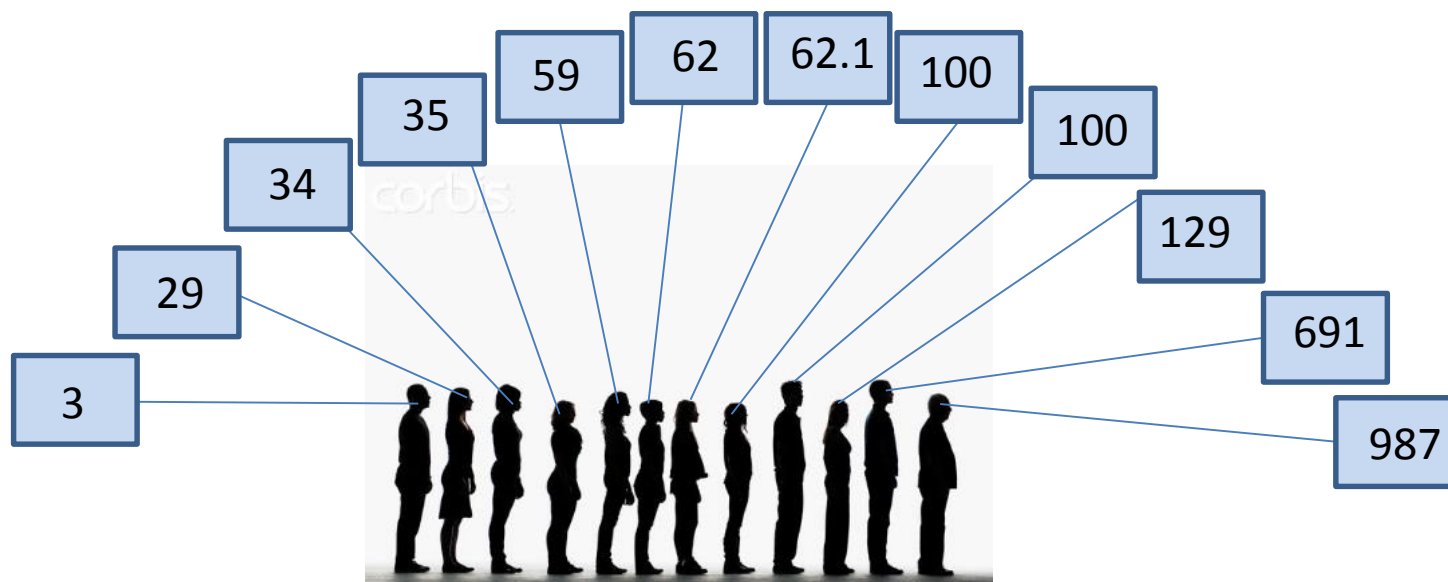


- Technically a “Last-In-First-Out Queue” or FIFO-Queue
- What happens:

Q.INSERT(A)
Q.INSERT(B)
Q.TOP()
Q.INSERT(C)
Q.POP()
Q.INSERT(D)
Q.POP()
Q.TOP()
Q.POP()
Q.INSERT(E)

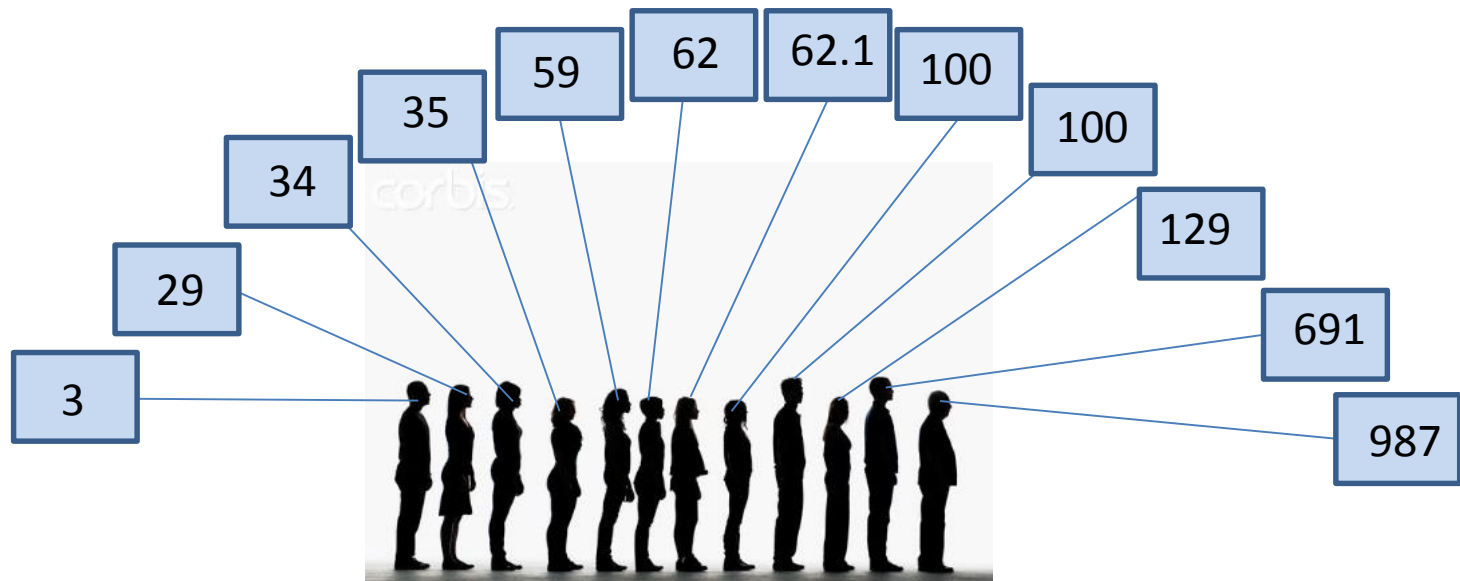
Q.INSERT(A)
Q.POP()
Q.POP()

Priority Queue (Priority Heap)



- Nodes are sorted by key values, $v.KEY$
 - Min-Priority Queue vs. Max-Priority Queue
- $Q.INSET(v)$ adds node v where it goes
- $Q.TOP()$ returns node with min/max key
- $Q.POP()$ returns & removes the node at “top”
- $Q.REMOVE(v)$ returns & removes node v
- $Q.UPDATE(v)$ updates position of v (after key change)

Priority Queue (Priority Heap)



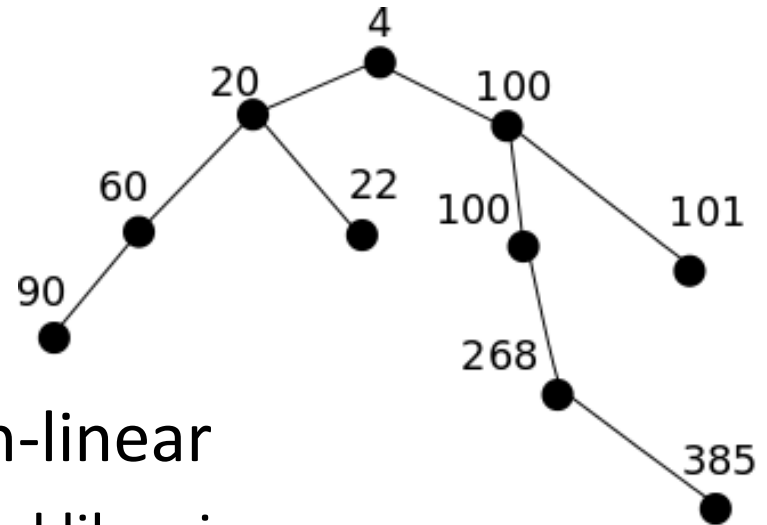
- Assume max-priority queue:

```
A.KEY = 5;      Q.INSERT(A);
B.KEY = 3;      Q.INSERT(B);
C.KEY = 10;     Q.INSERT(C);
B.KEY = 7;      Q.UPDATE(B);
Q.TOP();
A.KEY = 12;     Q.UPDATE(A);
```

```
Q.REMOVE(C);
Q.POP();
D.KEY = 4;      Q.INSERT(D);
B.KEY = 2;      Q.UPDATE(B);
Q.REMOVE(TOP());
Q.POP();
```

Priority Queue (Priority Heap)

- Why the name Heap?



- Fast data structures are non-linear
 - A good reason to use standard libraries....

	Binary	Binomial	Fibonacci	Brodal
TOP()	$O(1)$	$O(\log n)$	$O(1)$	$O(1)$
POP()	$O(\log n)$	$O(\log n)$	$O(\log n)$	$O(\log n)$
UPDATE(v)	$O(\log n)$	$O(\log n)$	$O(1)$	$O(1)$
INSERT(v)	$O(\log n)$	$O(1)$	$O(1)$	$O(1)$
REMOVE(v)	$O(\log n)$	$O(\log n)$	$O(\log n)$	$O(\log n)$

Graph Search

Simple Forward Graph Search Algorithms

- Breadth-First Search
 - Queue (FIFO) as subroutine
- Depth First Search
 - Stack (LIFO) as subroutine
- Best-First Search
 - Priority Queue as subroutine
 - Dijkstra's Algorithm
 - Heuristics can make “best” more accurate
 - A* Algorithm

Forward Graph Search: Preliminaries

- “start” node in V
- “goal” node in V
 - Sometimes we have multiple goals (don’t care which we get)
- Each node maintains a list of its neighbors
 - $v.\text{Neighbors} = \bigcup \{u \mid (v, u) \in E\}$
 - array or a list in practice
- $v.\text{NextNeighbor}()$
 - “iterator” function over $v.\text{Neighbors}$
 - each time we call this we get a new neighbor of v
 - after returning each neighbor once, it always returns \emptyset .
- $v.\text{parent}$
 - “parent pointer” that remembers search discovery order
 - the set of all parent pointers defines a **search-tree**!

Forward Graph Search: Preliminaries

- UNVISITED
 - a set that initially contains all nodes
 - nodes are removed when they are visited
- Algorithmic notation: “ \leftarrow ” vs. “ $=$ ”
 - “ \leftarrow ” assignment operator
 - $A \leftarrow B$
 - sets the value of A to be the same as that in B
 - “ $=$ ” boolean function that checks for equality
 - $A=B$
 - returns **true** if a and b have the same value
 - returns **false** if a and be have different values

Simple Forward Graph Search: Pseudocode

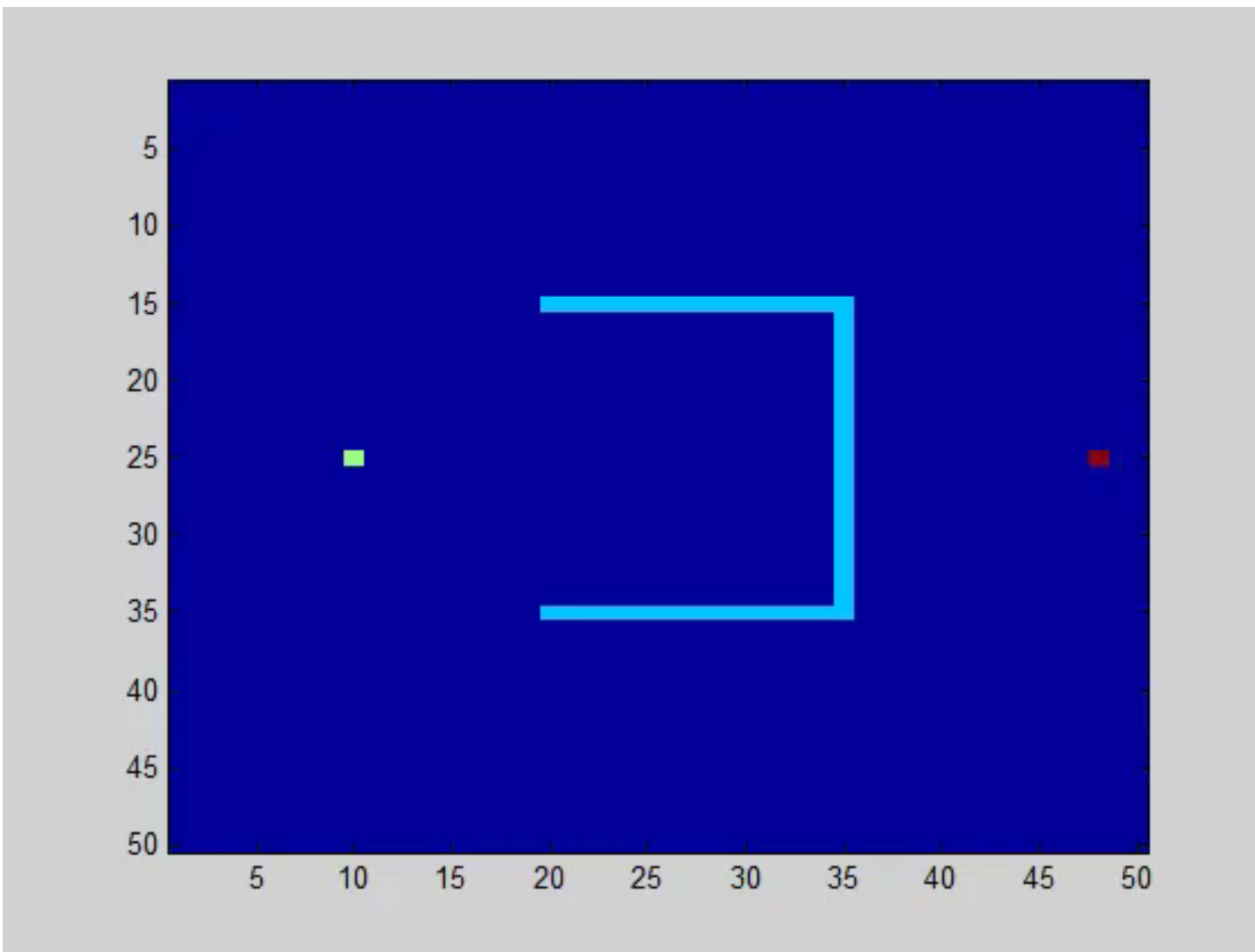
Function GraphSearch($G, \text{start}, \text{goal}, Q$)

1. $\text{UNVISITED} \leftarrow V \setminus \{\text{start}\}$
2. $Q.\text{INSERT}(\text{start})$
3. **while** $Q.\text{TOP}() \neq \emptyset$
4. $v \leftarrow Q.\text{TOP}()$
5. $u \leftarrow v.\text{NextNeighbor}()$
6. **if** $u = \emptyset$
7. $Q.\text{POP}()$
8. **else if** $u \in \text{UNVISITED}$
9. $\text{UNVISITED} \leftarrow \text{UNVISITED} \setminus \{u\}$
10. $u.\text{parent} \leftarrow v$
11. $Q.\text{INSERT}(u)$
12. **if** $u = \text{goal}$
13. **return** SUCCESS
14. **return** FAILURE

Breadth-First Search (BFS)

- Q is a FIFO-Queue
- The **search tree** (of parent pointers) goes as wide as possible before going deeper.
- Example on board

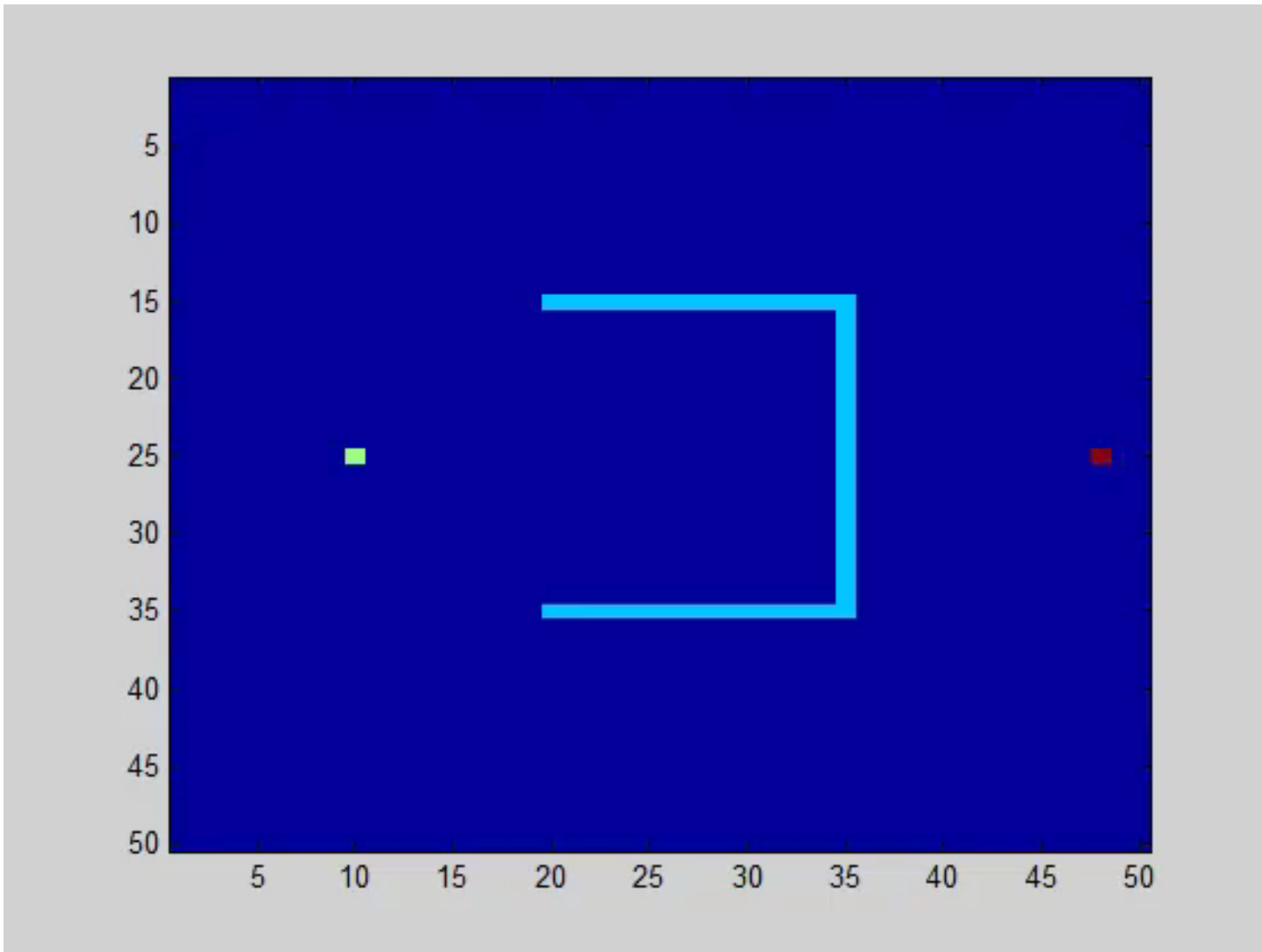
Breadth First Search (BFS): ANIMATION



Depth-First Search (DFS)

- Q is a Stack (LIFO-Queue)
- The **search tree** (of parent pointers) goes as deep as possible before going deeper.
- Example on board....

Depth First Search (DFS): ANIMATION



Best-First Search (BFS): More Preliminaries

- Q is a Priority Queue
- Only useful if we have some notion of “best”
 - Distance and cost are widely used in path/motion planning
 - “distance” e.g., Euclidian distance along trajectories
 - “cost” e.g., fuel required to go along trajectories
 - etc.
- $\text{Cost}(v,u)$
 - returns the cost of moving
 - from: the configuration space point represented by v
 - to: the configuration space point represented by u
 - Note, sometimes for $(u,v) \notin V$ we define $\text{Cost}(u,v) = \infty$

Best-First Search (BFS): More Preliminaries

- UNVISITED, OPEN, CLOSED
 - Sometimes this is UNVISITED, LIVE, DEAD
- UNVISITED
 - Node is not in search tree
 - (has not been “touched” yet)
- OPEN
 - Node is on the fringe of the search tree
 - It is currently in the priority queue
- CLOSED
 - Node is internal to the search tree
 - Node has been removed from the priority queue

Dijkstra's Pseudocode

Function DijkstraAlgorithm($G, start, goal, Q$)

1. UNVISITED $\leftarrow V \setminus \{start\}$
2. Q.INSERT(start)
3. **while** Q.TOP() $\neq \emptyset$ // and goal \notin CLOSED
4. $v \leftarrow Q.TOP()$
5. $u \leftarrow v.NextNeighbor()$
6. **if** $u = \emptyset$
7. Q.POP() //; OPEN \leftarrow OPEN $\setminus \{v\}$
//; CLOSED \leftarrow CLOSED $\cup \{v\}$
8. **else if** $u \in \text{Unvisited}$ **or** $u.costToStart > v.costToStart + \text{Cost}(v, u)$
9. UNVISITED \leftarrow UNVISITED $\setminus \{u\}$ //; OPEN \leftarrow OPEN $\cup \{u\}$
10. $u.parent \leftarrow v$
11. $u.costToStart = v.costToStart + \text{Cost}(v, u)$
12. Q.INSERT(u)
13. **if** $u = goal$
14. **return** SUCCESS
15. **return** FAILURE

Best-First Search: Dijkstra's

- Example on board....



Best-First Search: Dijkstra's

- So... hmm.... What's the difference between
- Depth-First Search and Dijkstra's algorithm?



Bibliography for this Lecture

- LaValle, S. M. Planning Algorithms, 2006