



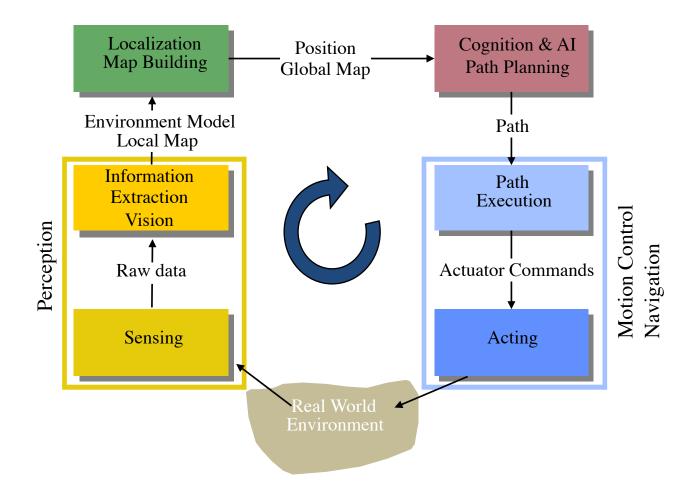
CS283: Robotics Fall 2020: Path Planning

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ShanghaiTech University

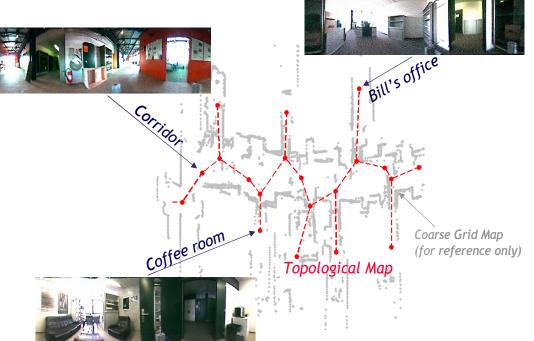
PLANNING

General Control Scheme for Mobile Robot Systems



The Planning Problem

- The problem: find a path in the work space (physical space) from the initial position to the goal position avoiding all collisions with the obstacles
- Assumption: there exists a good enough map of the environment for navigation.

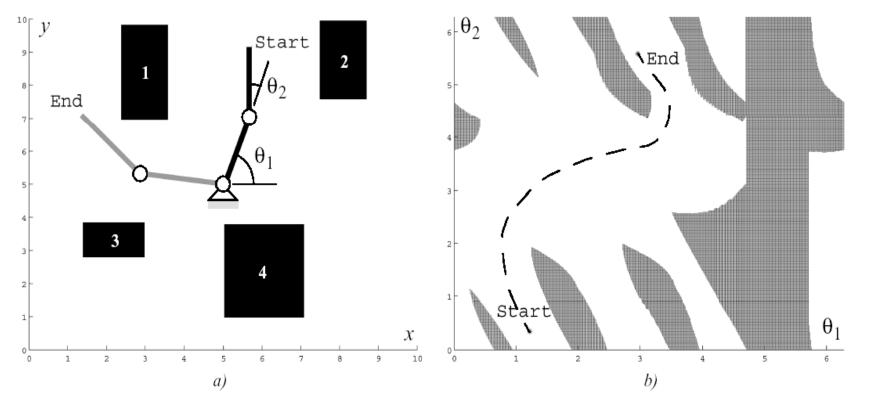


The Planning Problem

- We can generally distinguish between
 - (global) path planning and
 - (local) obstacle avoidance.
- First step:
 - Transformation of the map into a representation useful for planning
 - This step is planner-dependent
- Second step:
 - Plan a path on the transformed map
- Third step:
 - Send motion commands to controller
 - This step is planner-dependent (e.g. Model based feed forward, path following)

Work Space (Map) → Configuration Space

• State or configuration *q* can be described with *k* values *q_i*



Work Space

• What is the configuration space of a mobile robot?

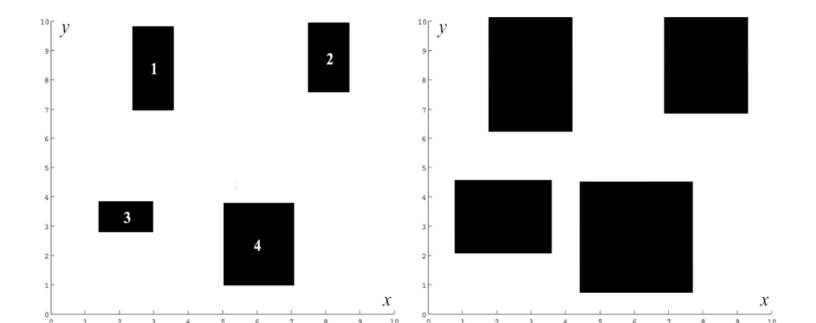
Configuration Space: the dimension of this

space is equal to the Degrees of Freedom (DoF) of the robot

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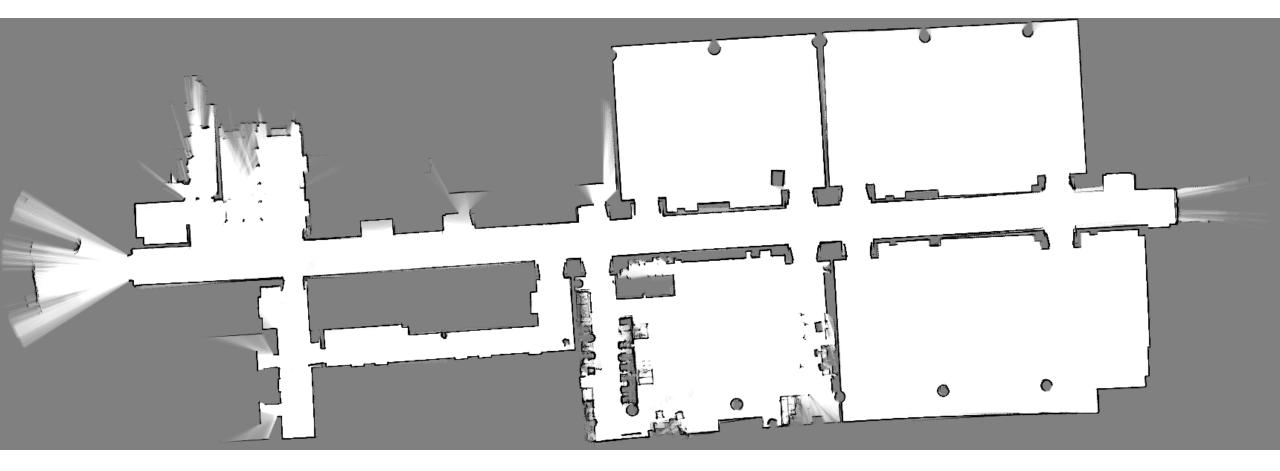
Configuration Space for a Mobile Robot

- Mobile robots operating on a flat ground (2D) have 3 DoF: (x, y, θ)
- Differential Drive: only two motors => only 2 degrees of freedom directly controlled (forward/ backward + turn) => non-holonomic
- Simplification: assume robot is holonomic and it is a point => configuration space is reduced to 2D (x,y)
- => inflate obstacle by size of the robot radius to avoid crashes => obstacle growing



Typical Configuration Space: Occupancy grid

• Fixed cell decomposition: occupancy grid example: STAR Center



Path Planning: Overview of Algorithms

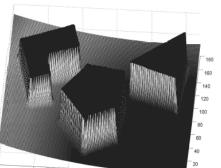
1. Optimal Control

- Solves truly optimal solution
- Becomes intractable for even moderately complex as well as nonconvex problems

Source: http://mitocw.udsm.ac.tz

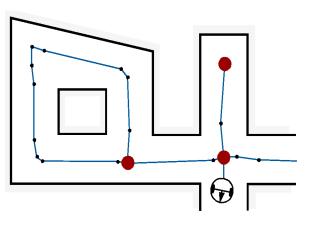
2. Potential Field

- Imposes a mathematical function over the state/configuration space
- Many physical metaphors exist
- Often employed due to its simplicity and similarity to optimal control solutions

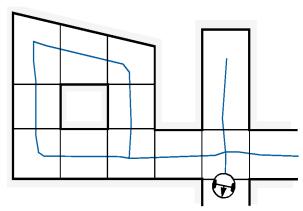


3. Graph Search

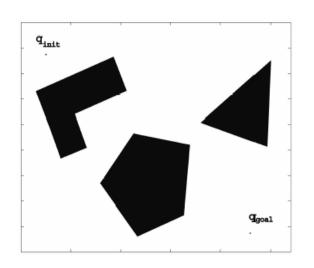
Identify a set edges between nodes within the free space



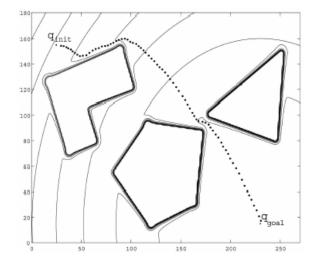
• Where to put the nodes?

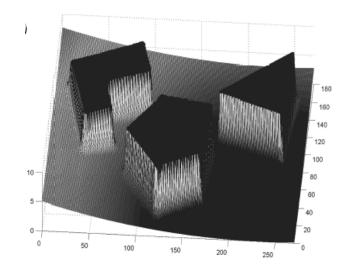


Potential Field Path Planning Strategies



- Robot is treated as a *point under the influence* of an artificial potential field.
- Operates in the continuum
 - Generated robot movement is similar to a ball rolling down the hill
 - Goal generates attractive force
 - Obstacle are repulsive forces

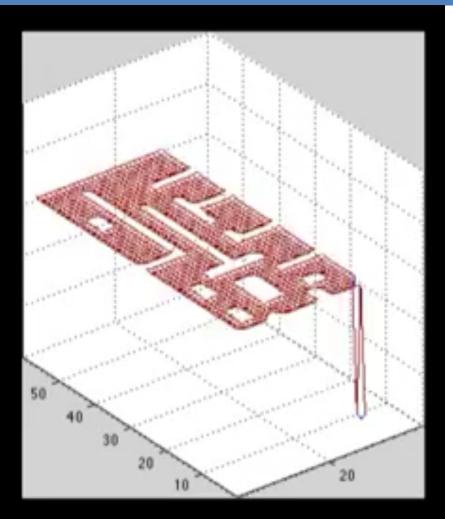












Robot Path Planning and Obstacle Avoidance using Harmonic Potential Fields

 $\begin{bmatrix} \partial U \end{bmatrix}$

Potential Field Path Planning: Potential Field Generation

• Generation of potential field function U(q)

- attracting (goal) and repulsing (obstacle) fields
- summing up the fields
- functions must be differentiable
- Generate artificial force field F(q)

$$F(q) = -\nabla U(q) = -\nabla U_{att}(q) - \nabla U_{rep}(q) = \begin{bmatrix} \frac{\partial U}{\partial x} \\ \frac{\partial U}{\partial y} \end{bmatrix}$$

- Set robot speed (v_x, v_y) proportional to the force F(q) generated by the field
 - the force field drives the robot to the goal
 - if robot is assumed to be a point mass
 - Method produces both a plan and the corresponding control

Potential Field Path Planning: Attractive Potential Field

• Parabolic function representing the Euclidean distance $\rho_{goal} = ||q - q_{goal}||$ to the goal

$$U_{att}(q) = \frac{1}{2}k_{att} \cdot \rho_{goal}^2(q)$$
$$= \frac{1}{2}k_{att} \cdot (q - q_{goal})^2$$

Attracting force converges linearly towards 0 (goal)

$$F_{att}(q) = -\nabla U_{att}(q)$$
$$= k_{att} \cdot (q - q_{goal})$$

Potential Field Path Planning: Repulsing Potential Field

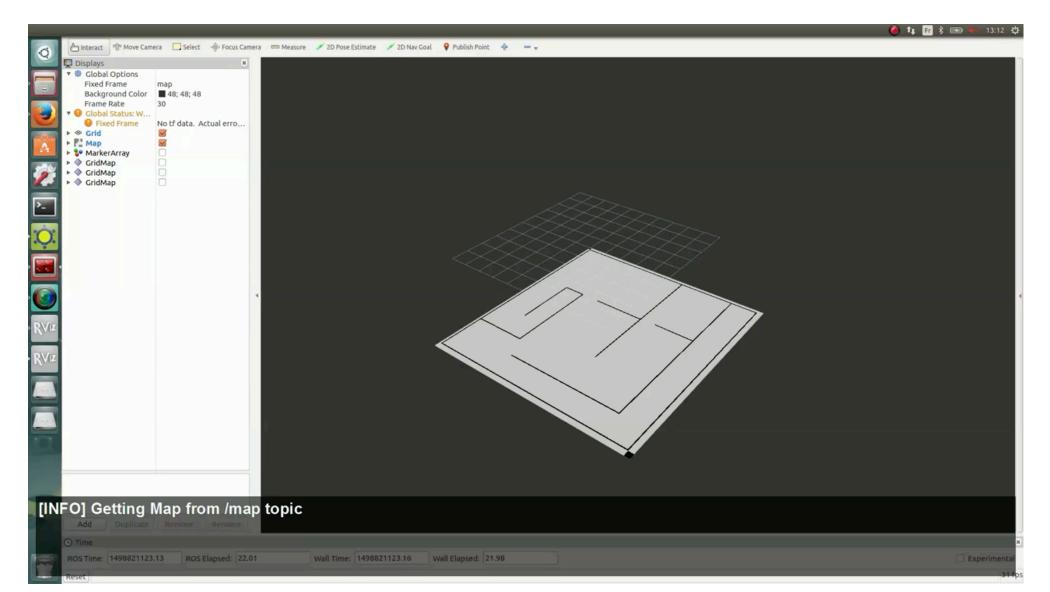
- Should generate a barrier around all the obstacle
 - strong if close to the obstacle
 - not influence if far from the obstacle

$$U_{rep}(q) = \begin{cases} \frac{1}{2} k_{rep} \left(\frac{1}{\rho(q)} - \frac{1}{\rho_0} \right)^2 & \text{if } \rho(q) \le \rho_0 \\ 0 & \text{if } \rho(q) \ge \rho_0 \end{cases}$$

- $\rho(q)$: minimum distance to the object
- Field is positive or zero and *tends to infinity* as q gets closer to the object

ROS Grid Map Package

http://wiki.ros.org/grid_map



Potential Field Path Planning:

• Notes:

- Local minima problem exists
- Problem is getting more complex if the robot is not considered as a point mass
- If objects are non-convex there exists situations where several minimal distances exist \rightarrow can result in oscillations

Example Configuration Space



Potential Field Path Planning: Extended Potential Field Method

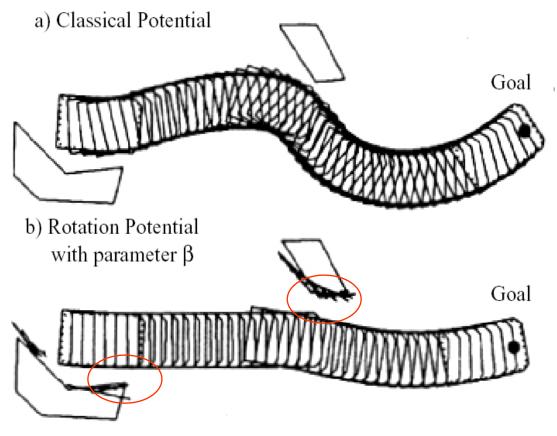
 Additionally a rotation potential field and a task potential field is introduced

Rotation potential field

 force is also a function of robots orientation relative to the obstacles. This is done using a gain factor that reduces the repulsive force when obstacles are parallel to robot's direction of travel

Task potential field

 Filters out the obstacles that should not influence the robots movements, i.e. only the obstacles in the sector in front of the robot are considered

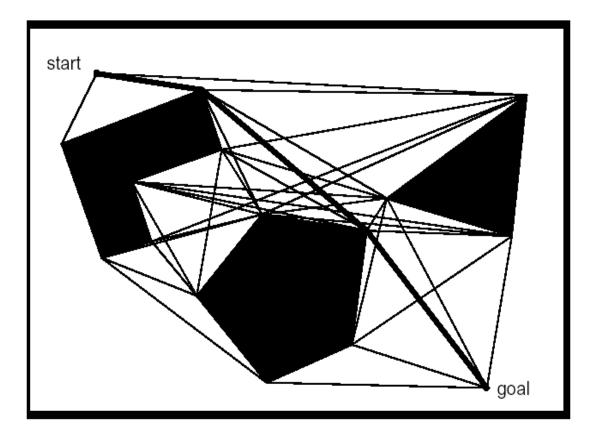


Khatib and Chatila

Graph Search

- Overview
 - Solves a least cost problem between two states on a (directed) graph
 - Graph structure is a discrete representation
- Limitations
 - State space is discretized \rightarrow completeness is at stake
 - Feasibility of paths is often not inherently encoded
- Algorithms
 - (Preprocessing steps)
 - Breath first
 - Depth first
 - Dijkstra
 - A* and variants
 - D* and variants

Graph Construction: Visibility Graph



- Particularly suitable for polygon-like obstacles
- Shortest path length
- Grow obstacles to avoid collisions

Graph Construction: Visibility Graph

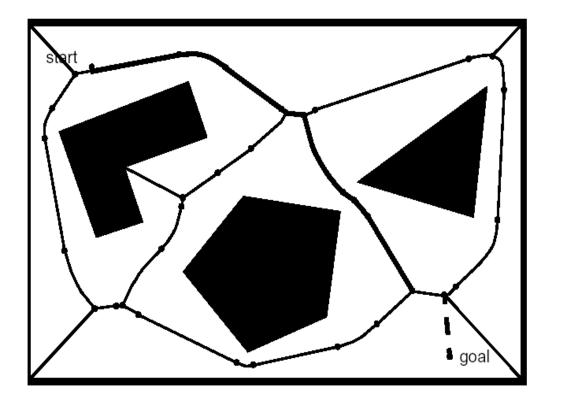
Pros

- The found path is optimal because it is the shortest length path
- Implementation simple when obstacles are polygons

Cons

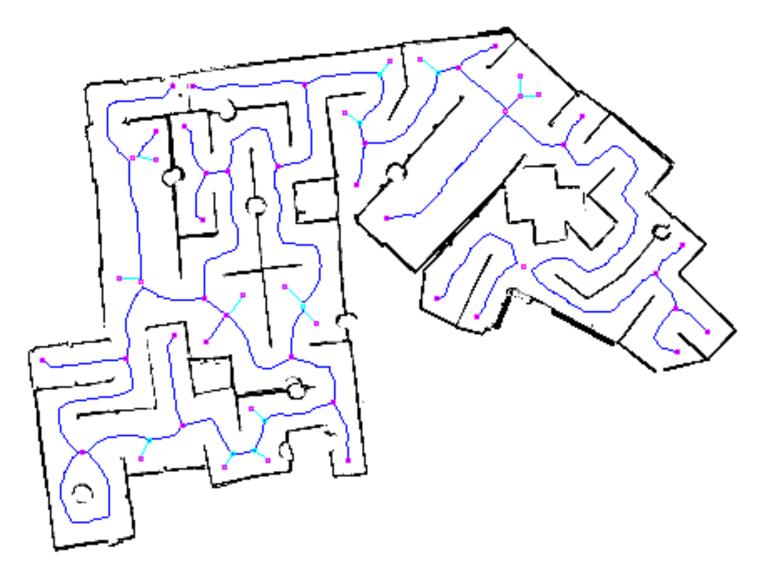
- The solution path found by the visibility graph tend to take the robot as close as possible to the obstacles: the common solution is to grow obstacles by more than robot's radius
- Number of edges and nodes increases with the number of polygons
- Thus it can be inefficient in densely populated environments

Graph Construction: Voronoi Diagram



Tends to maximize the distance between robot and obstacles

Topology Graph



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Graph Construction: Voronoi Diagram

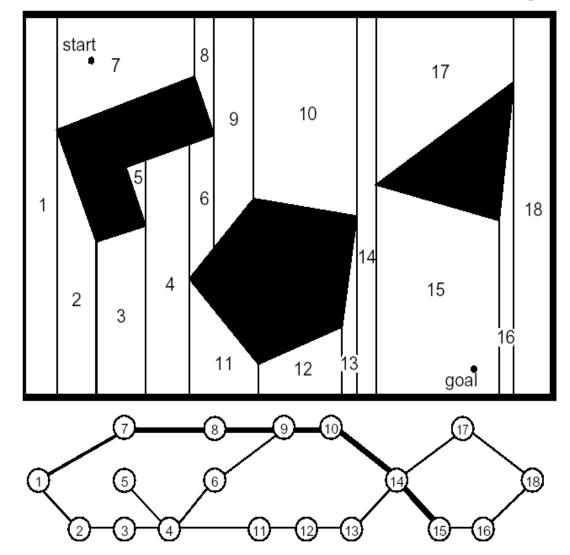
Pros

 Using range sensors like laser or sonar, a robot can navigate along the Voronoi diagram using simple control rules

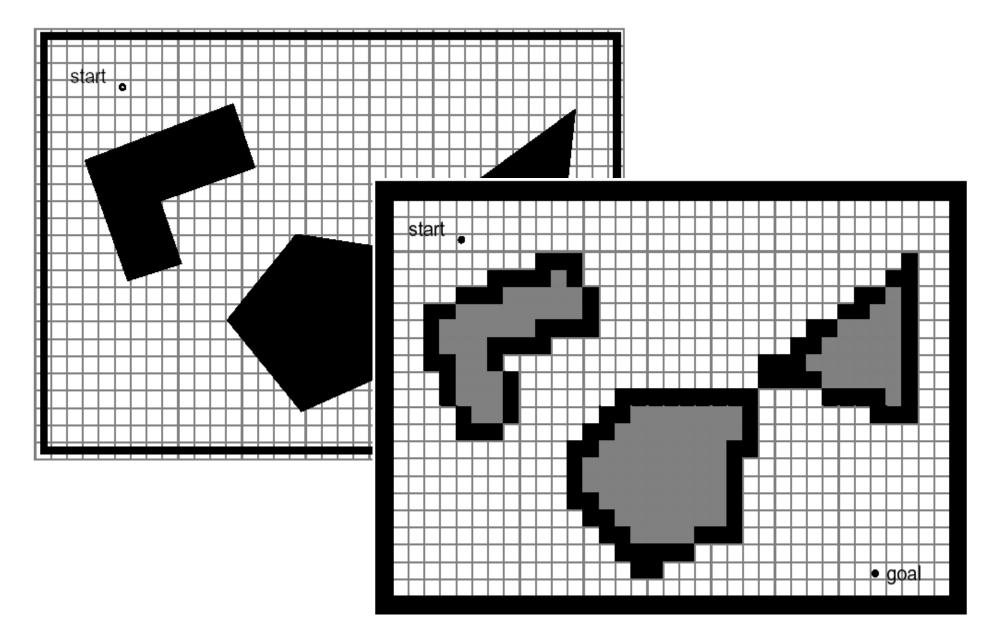
Cons

- Because the Voronoi diagram tends to keep the robot as far as possible from obstacles, any short range sensor will be in danger of failing
- Voronoi diagram can change drastically in open areas

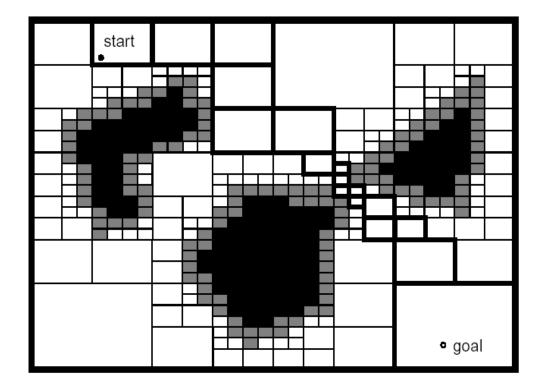
Graph Construction: Exact Cell Decomposition (2/4)



Graph Construction: Approximate Cell Decomposition (3/4)



Graph Construction: Adaptive Cell Decomposition (4/4)

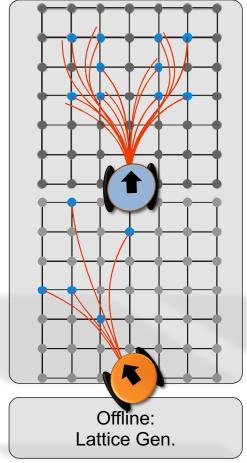


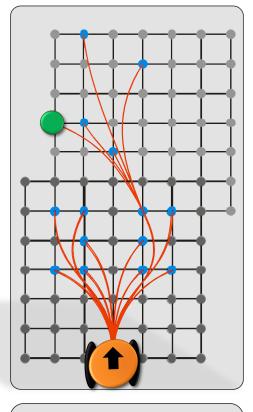
Close relationship with map representation (Quadtree)!

Graph Construction: State Lattice Design (1/2)

Enforces edge feasibility





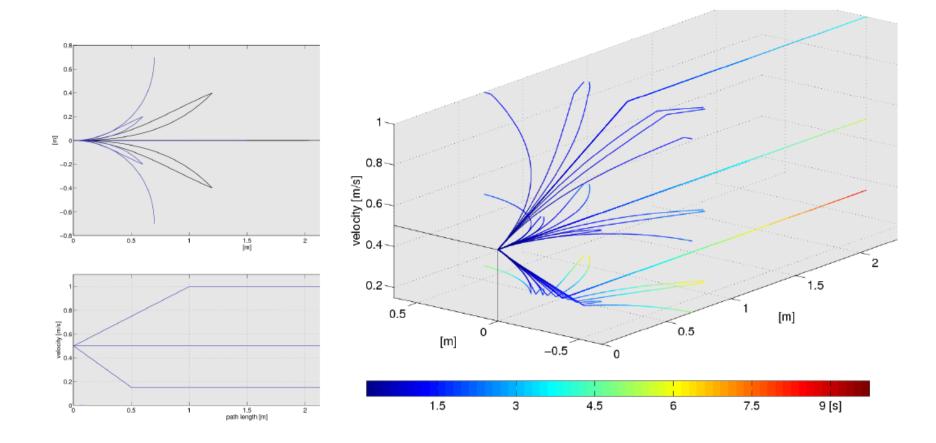


Online: Incremental Graph Constr.

Graph Construction: State Lattice Design (2/2)

Martin Rufli

State lattice encodes only kinematically feasible edges

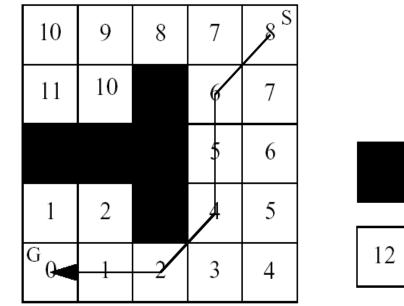


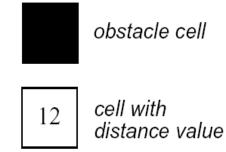
Deterministic Graph Search

- Methods
 - Breath First
 - Depth First
 - Dijkstra

•

- A* and variants
- D* and variants





ADMIN

Admin

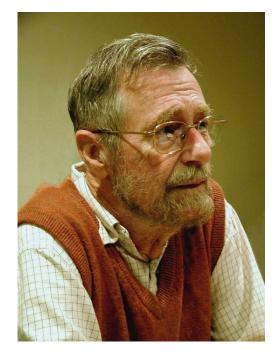
Paper presentation ppt/ pdf due today!

• HW3:

- Bug was found regarding the mirrored maps!
- Check for the solution on piazza.

DIJKSTRA'S ALGORITHM

EDSGER WYBE DIJKSTRA



1930 - 2002

"Computer Science is no more about computers than astronomy is about telescopes."

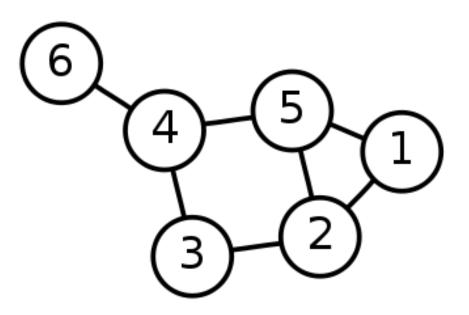
http://www.cs.utexas.edu/~EWD/

SINGLE-SOURCE SHORTEST PATH PROBLEM

 <u>Single-Source Shortest Path Problem</u> - The problem of finding shortest paths from a source vertex v to all other vertices in the graph.

Graph

- Set of vertices and edges
- Vertex:
 - Place in the graph; connected by:
- Edge: connecting two vertices
 - Directed or undirected (undirected in Dijkstra's Algorithm)
 - Edges can have weight/ distance assigned



Diklstra's Algorithm

- Assign all vertices infinite distance to goal
- Assign 0 to distance from start
- Add all vertices to the queue
- While the queue is not empty:
 - Select vertex with smallest distance and remove it from the queue
 - Visit all neighbor vertices of that vertex,
 - calculate their distance and
 - update their (the neighbors) distance if the new distance is smaller

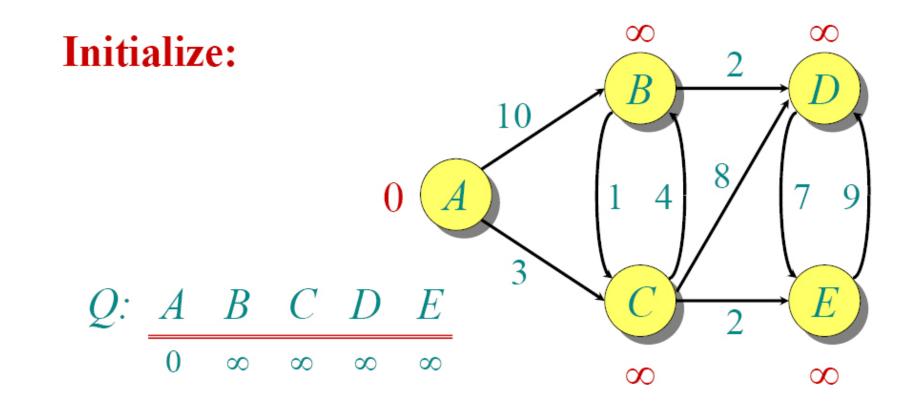
Diklstra's Algorithm - Pseudocode

```
dist[s] \leftarrow o
for all v \in V - \{s\}
      do dist[v] \leftarrow \infty
S \leftarrow \emptyset
O← V
while Q ≠Ø
do u \leftarrow mindistance(Q, dist)
    S \leftarrow S \cup \{u\}
     for all v \in neighbors[u]
          do if dist[v] > dist[u] + w(u, v)
         (if desired, add traceback code)
return dist
```

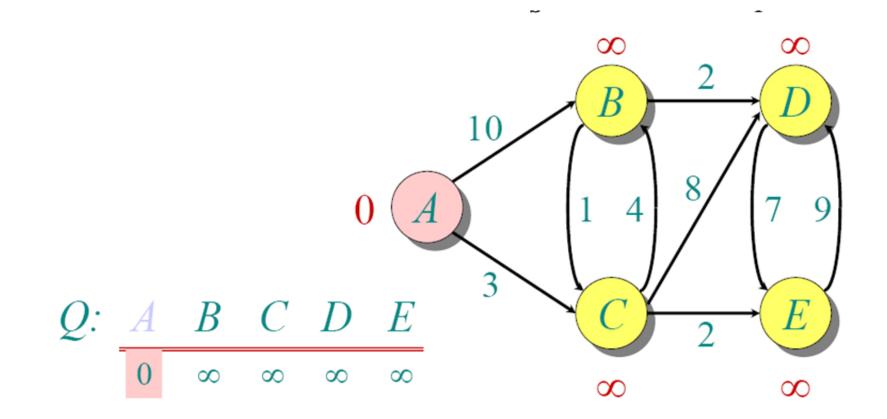
(distance to source vertex is zero)

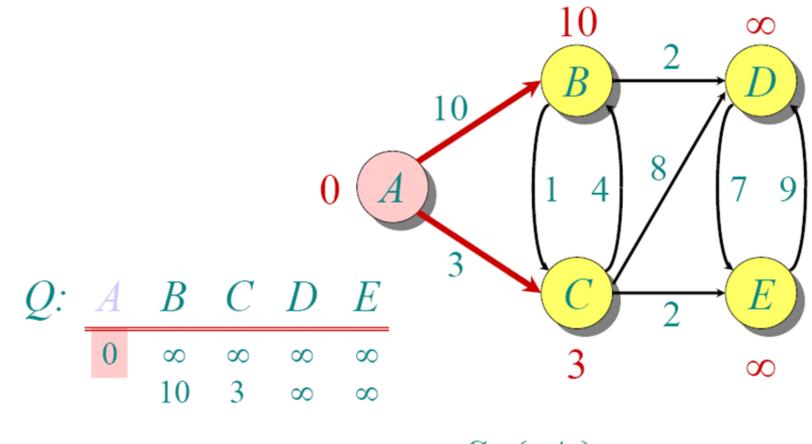
(set all other distances to infinity)
(S, the set of visited vertices is initially empty)
(Q, the queue initially contains all vertices)
(while the queue is not empty)
(select the element of Q with the min. distance)
(add u to list of visited vertices)

 $\begin{array}{ll} dist[v] > dist[u] + w(u, v) & (if new shortest path found) \\ then & d[v] \leftarrow d[u] + w(u, v) & (set new value of shortest path) \end{array}$

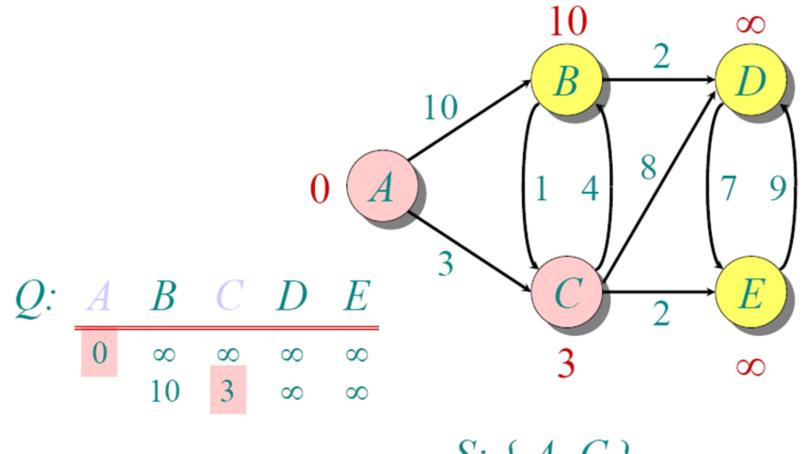


S: {}

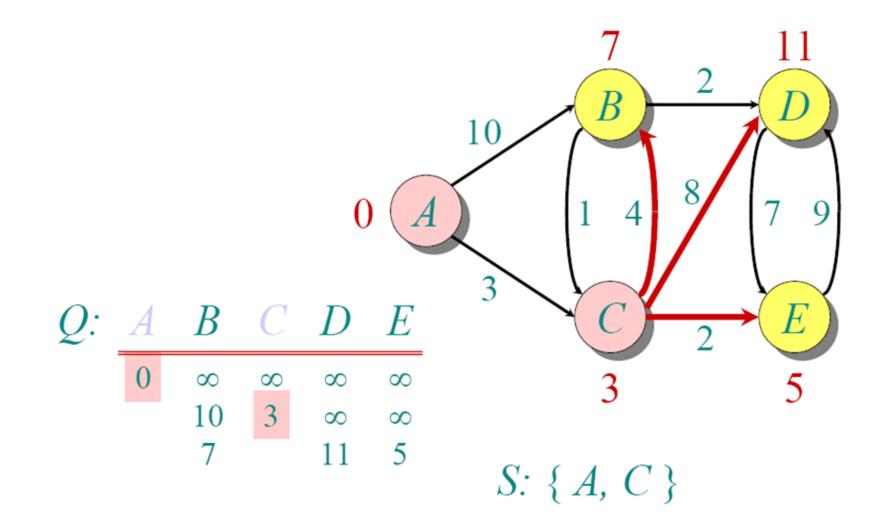


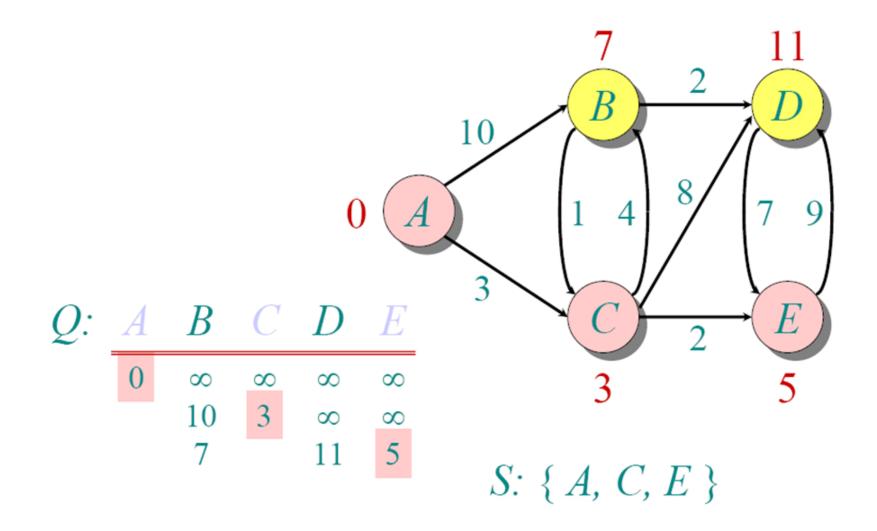


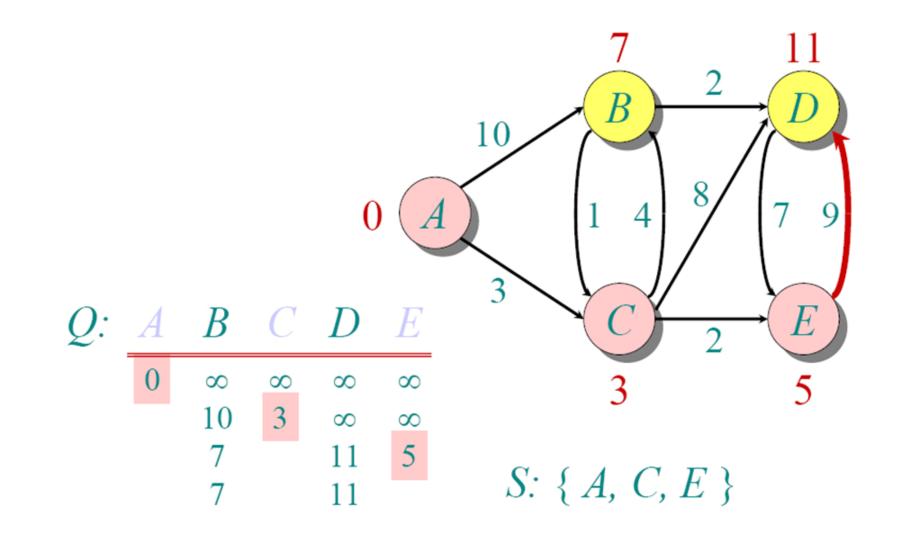
 $S: \{A\}$

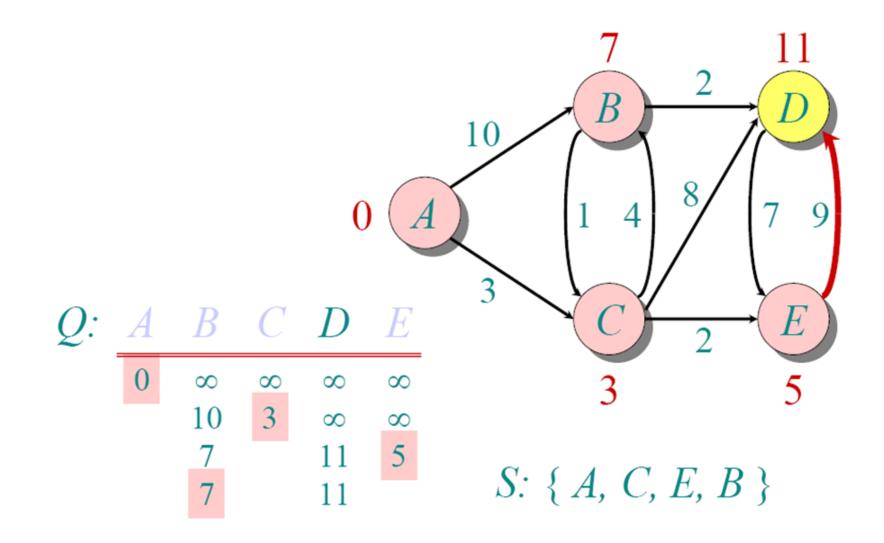


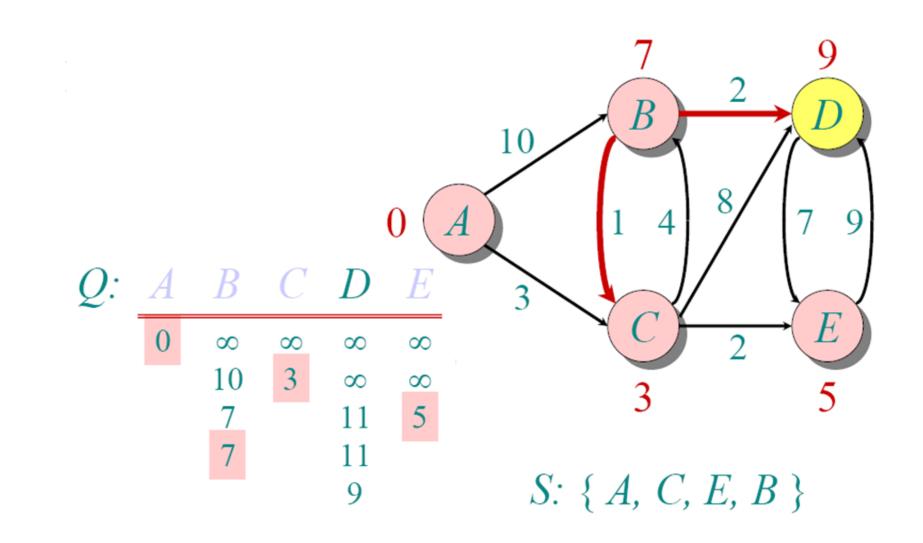
 $S: \{A, C\}$

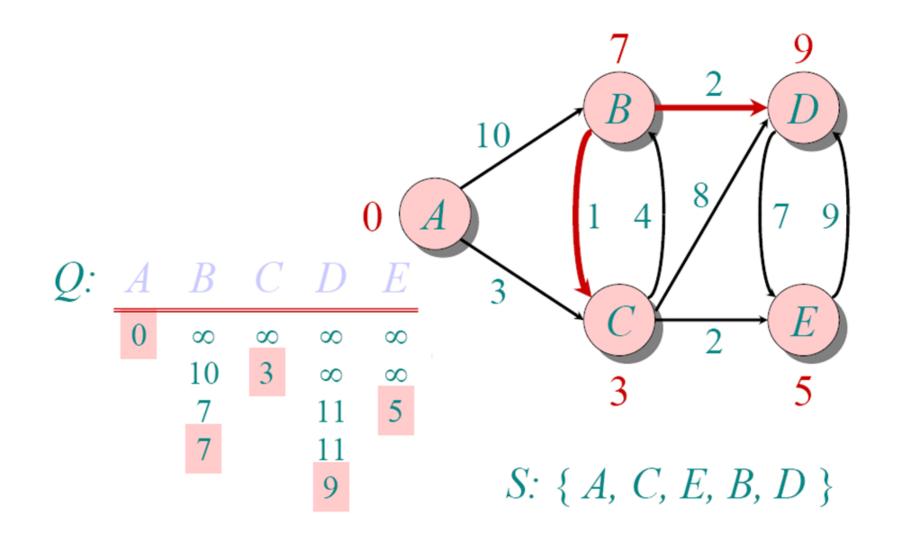










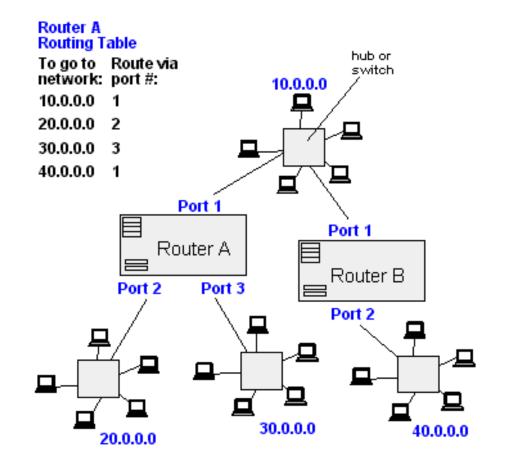


APPLICATIONS OF DIJKSTRA'S ALGORITHM

- Navigation Systems
- Internet Routing

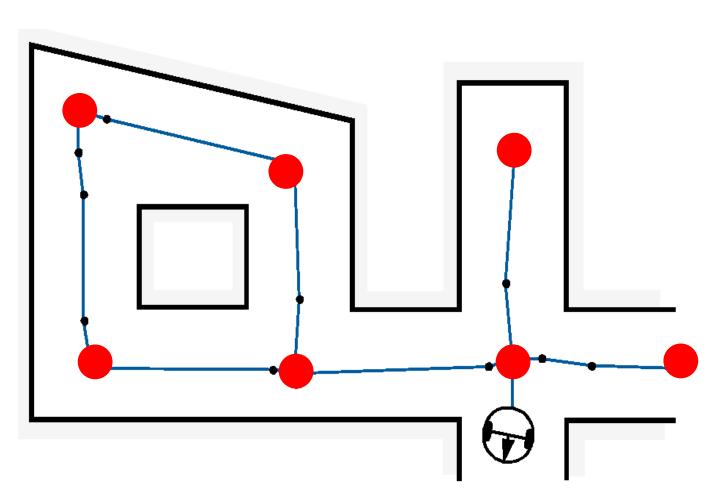


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Dijkstra's Algorithm for Path Planning: Topological Maps

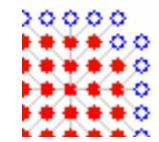
- Topological Map:
 - Places (vertices) in the environment (red dots)
 - Paths (edges) between them (blue lines)
 - Length of path = weight of edge
- => Apply Dijkstra's Algorithm to find path from start vertex to goal vertex



Dijkstra's Algorithm for Path Planning: Grid Maps

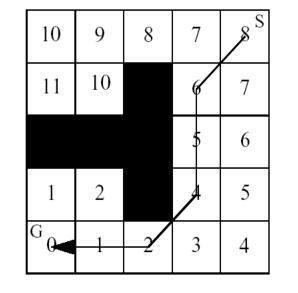
• Graph:

- Neighboring free cells are connected:
 - 4-neighborhood: up/ down/ left right
 - 8-neighborhood: also diagonals
- All edges have weight 1
- Stop once goal vertex is reached
- Per vertex: save edge over which the shortest distance from start was reached => Path



Graph Search Strategies: Breath-First Search

- Corresponds to a wavefront expansion on a 2D grid
- Breath-First: Dijkstra's search where all edges have weight 1

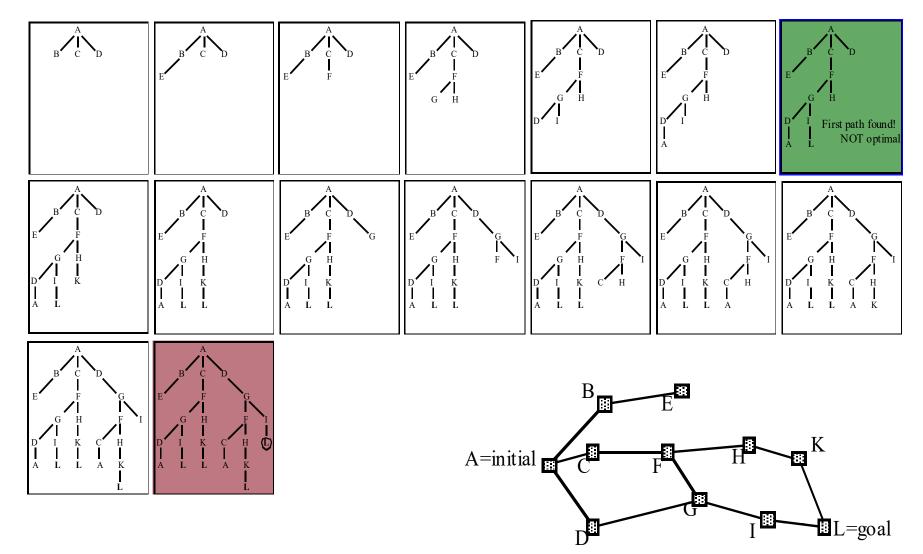




cell with distance value

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Graph Search Strategies: Depth-First Search



Graph Search Strategies: A* Search

- Similar to Dijkstra's algorithm, except that it uses a heuristic function h(n)
- f(n) = g(n) + h(n)

anal		g=1.4	g=1.0	cool		g=1.4	g=1.0	anal		g=1.4	g=1.0	anal		g=1.4	g=1.0
goal		h=2.0	h=3.0	goal		h=2.0	h=3.0	goal		h=2.0	h=3.0	goal		h=2.0	h=3.0
			start				start				start				start
		g=1.4	g=1.0			g=1.4	g=1.0			g=1.4	g=1.0		g=2.4	g=1.4	g=1.0
		h=2.8	h=3.8			h=2.8	h=3.8			h=2.8	h=3.8		h=2.4	h=2.8	h=3.8
													g=2.8	g=2.4	g=2.8
													h=3.4	h=3.8	h=4.2
goal		g=1.4 h=2.0	g=1.0 h=3.0	g=4.8 goal h=0.0		g=1.4 h=2.0	g=1.0 h=3.0	g=4.8 goal h=0.0		g=1.4 h=2.0	g=1.0 h=3.0	goal			
g=3.8 h=1.0		11-2.0	start	g=3.8 h=1.0		11-2.0	start	g=3.8 h=1.0		11-2:0	start	L			start
g=3.4	g=2.4	g=1.4	g=1.0	g=3.4	g=2.4	g=1.4	g=1.0	g=3.4	g=2.4	g=1.4	g=1.0		$\overline{}$		
h=2.0	h=2.4	h=2.8	h=3.8	h=2.0	h=2.4	h=2.8	h=3.8	h=2.0	h=2.4	h=2.8	h=3.8				
g=3.8	g=2.8	g=2.4	g=2.8	g=3.8	g=2.8	g=2.4	g=2.8	g=3.8	g=2.8	g=2.4	g=2.8				
h=3.0	h=3.4	h=3.8	h=4.2	h=3.0	h=3.4	h=3.8	h=4.2	h=3.0	h=3.4	h=3.8	h=4.2				

A*

- Developed 1986 as part of the Shakey project!
- Complexity:

Worst-case performance $O(|E|) = O(b^d)$ Worst-case space $O(|V|) = O(b^d)$ complexity

- b: branching factord: depth
- Good heuristic => small branching factor

Optimal Planning

- Dijkstra finds the optimal path
- What about A*?
 - Find admissible heuristic h(n)
 - Admissible: do not overestimate the true cost-to-go
 - A* is optimal (finds optimal/ shortest path) if h(n) is admissible for all n
 - Admissible example: use geometric distance for h(n):

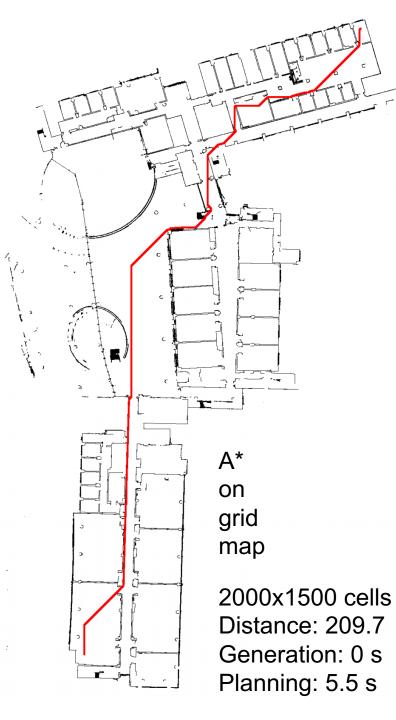
$$h(n) = \sqrt{(x_{goal} - x)^2 + (y_{goal} - y)^2}$$
).

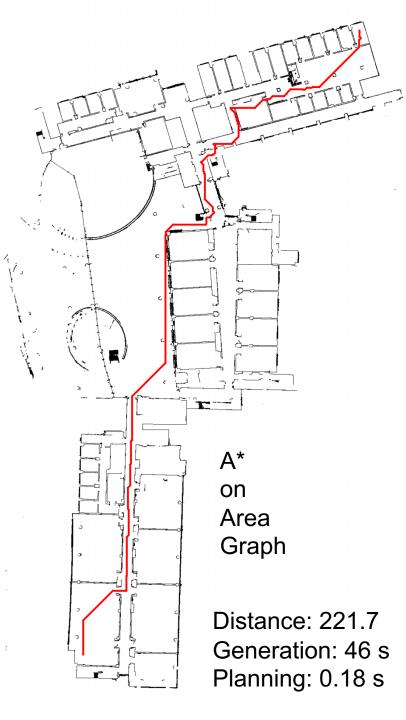
- Example: heuristic 5x geometric distance
- h(n) = 0 => Dijkstra's Algorithm

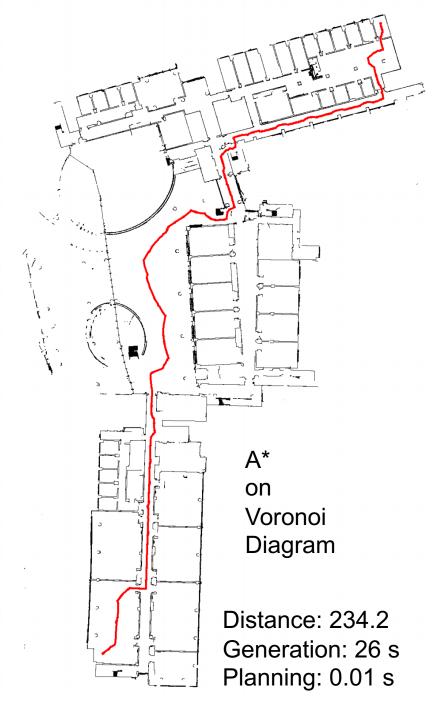
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A*

- Hierarchical planning possible e.g.: Go to the library:
 - First plan how to get from SIST building to library
 - Then plan how to get from entrance of library to goal room (campus level vs. library level)
- Many variants of A* algorithms exist with different properties
- A* as graph search: applications outside of robotics/ path planning
 - Video games
 - Parsing with stochastic grammars in natural language processing
- Graph on which planning is done matters!
 - E.g.: Grid map; Pose Graph; Topological Graph; Open Street Map; Lattice Graphs; ...

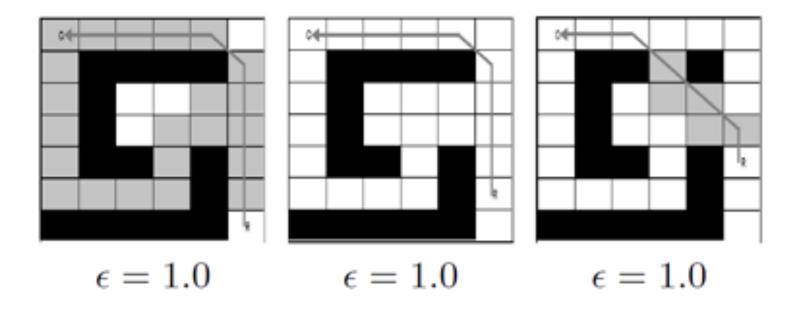






Graph Search Strategies: D* Search

- Similar to A* search, except that the search starts from the goal outward
- $f(n) = g(n) + \epsilon h(n)$
- First pass is identical to A*
- Subsequent passes reuse information from previous searches



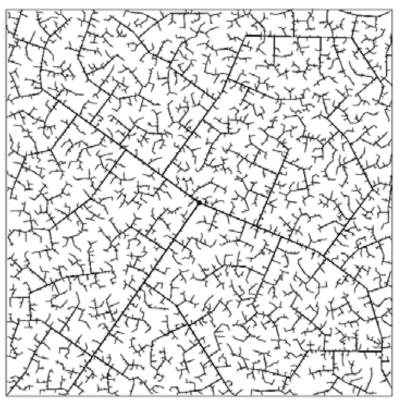


Graph Search Strategies: Randomized Search

- Most popular version is the rapidly exploring random tree (RRT)
 - Well suited for high-dimensional search spaces
 - Often produces highly suboptimal solutions



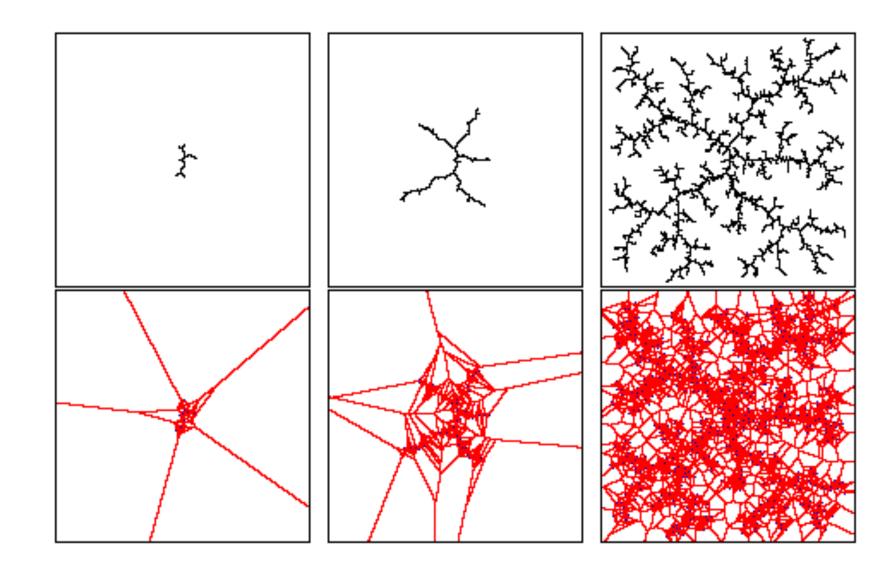
45 iterations



2345 iterations

Why are RRT's rapidly exploring?

The probability of a node to be selected for expansion is proportional to the area of its Voronoi region



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rtstar.h



rrtstarplan.cpp

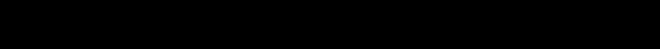


rrtstarplan.h

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goal: 5.96963 7.10589

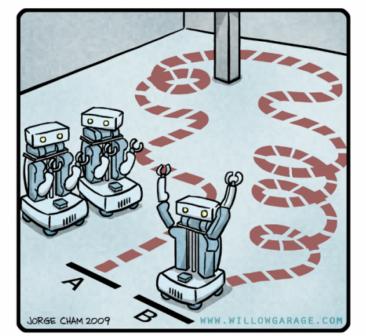
New Path Found. Total paths 1 Finding Optimal Path [INFO] [1497922923.557025739, 592.300000000]: Got new plan INFO] [1497922923.957553192, 592.700000000]: Goal reached ^C[rviz-13] killing on exit [amcl-12] killing on exit [map_server-11] killing on exit [move_base-10] killing on exit [kobuki_safety_controller-9] killing on exit [navigation_velocity_smoother-8] killing on exit [cmd_vel_mux-7] killing on exit [mobile_base_nodelet_manager-6] killing on exit [joint_state_publisher-5] killing on exit [diagnostic_aggregator-4] killing on exit [robot_state_publisher-3] killing on exit [stageros-2] killing on exit [rosout-1] killing on exit [master] killing on exit shutting down processing monitor... ... shutting down processing monitor complete done lizhi@lizhi-HP-EliteBook-8460w:~/download codes/rrtstar_planner\$



R.O.B.O.T. Comics

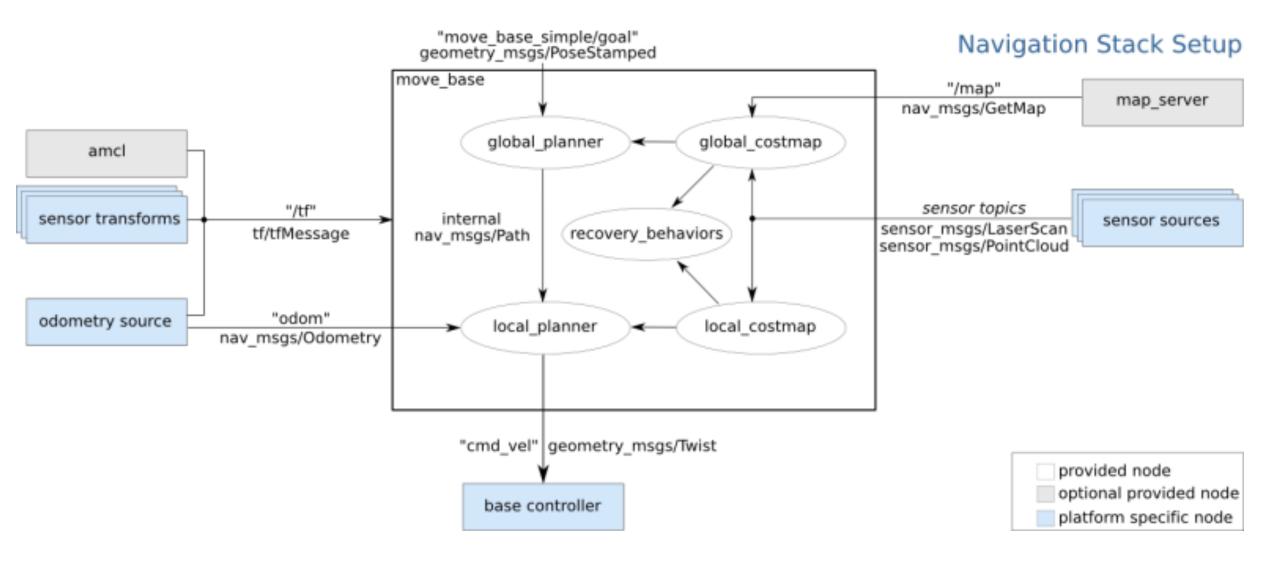
ROS Navigation

http://wiki.ros.org/navigation



"HIS PATH-PLANNING MAY BE SUB-OPTIMAL, BUT IT'S GOT FLAIR."

Path Planning in ROS: move_base







 $\mathbf{H}\mathbf{R}$

teb_local_planner

An optimal trajectory planner for mobile robots based on Timed-Elastic-Bands

http://wiki.ros.org/teb_local_planner