

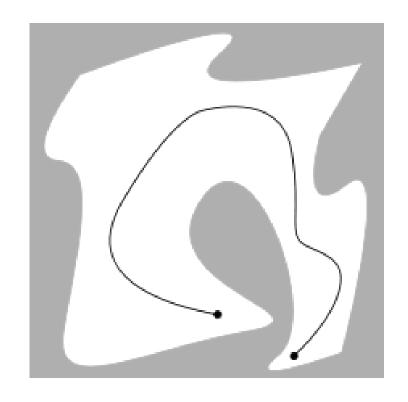
# **Motion Planning**

in-depth study

What is motion planning?

Define a sequence of actions to take a robot from one known state (initial) to another known state (goal), while:

- avoiding obstacles;
- respecting motion constraints;
- optimizing a cost function (hopefully)



#### **Assumptions**

- Ideal knowledge of the robot and the world
- A priori knowledge of actual and goal positions
- motion rules of our robot are fixed

#### Popular approaches:

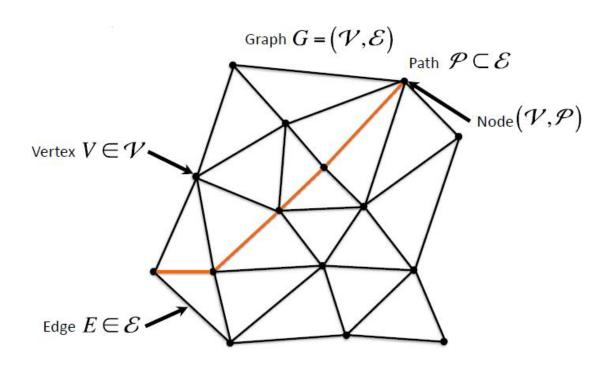
- Potential fields [Rimon, Koditschek, '92]: create forces on the robot that pull it toward the goal and push it away from obstacles
- Grid-based planning [Stentz, '94]: discretizes problem into grid and runs a graph-search algorithm (Dijkstra, A\*, ...)
- Combinatorial planning [LaValle, '06]: constructs structures in the configuration (C-) space that completely capture all information needed for planning
- Sampling-based planning [Kavraki et al, '96; LaValle, Kuffner, '06, etc.]: uses collision detection algorithms to probe and incrementally search the C-space for a solution, rather than completely characterizing all of the  $C_{\rm free}$  structure

#### Discrete graph representation

It is common to convert the planning problem into a (discrete) **graph** representation use one of the existing search algorithms on the graph

#### Where *edges* can:

- Have a direction (directed graph)
- Have a cost (weighted graph)



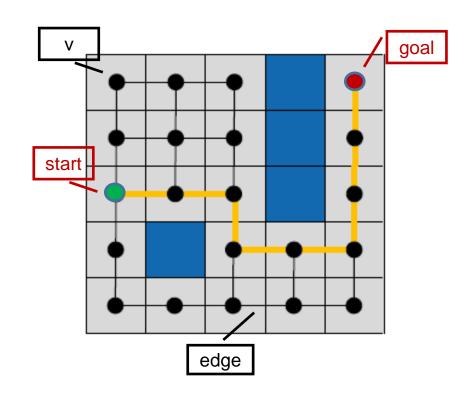
#### Grid based approach

#### **Discretize** the continuous world **into a grid**:

- Each cell is either free or forbidden
- Robot moves between adjacent free cells
- Goal: find sequence of free cells from start

Mathematically, this corresponds to pathfinding in a discrete graph G = (V, E)

- Each vertex  $v \in V$  represents a free cell
- Edges  $(v, u) \in E$  connect adjacent grid cells

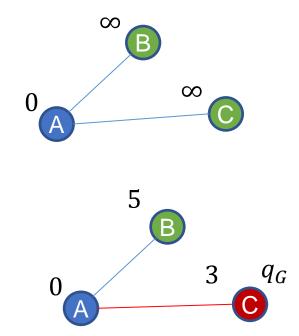


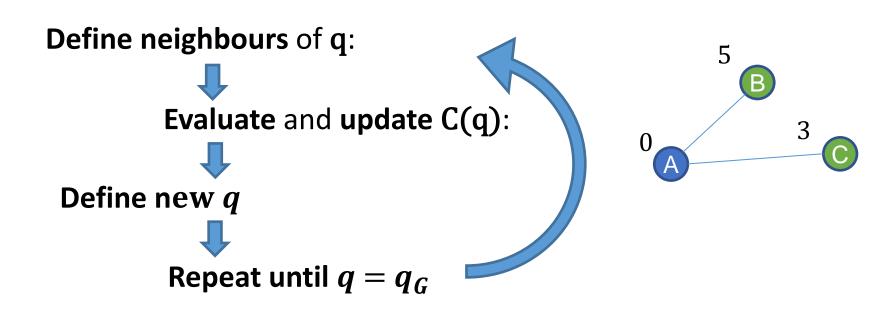
Define **Graph** and a possible **research "tree"** G

Graph search algorithms

### **Label correcting algorithms:**

best path optimizing cost-of-arrival C(q)  $q_I \text{ to } q$ 





### **Label correcting algorithms:**

best path optimizing cost-of-arrival C(q)  $q_I$  to q

Initialize  $C(nodes) = \infty$ :

**Remove q** from queue and **find q**':

**Evaluate** and **update** C(q'):

if lower set q as q' parent

if 
$$q' != q_G$$

Add q' in queue

Get next (q)



terminate

#### How to get next node?

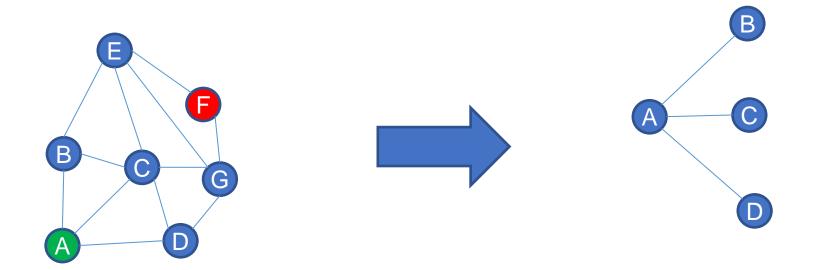
Breadth-First-Search (BFS, Bellman-Ford): Maintain Q as a list – First in/first out

Q: frontier/open/alive set of nodes = next nodes to be visited

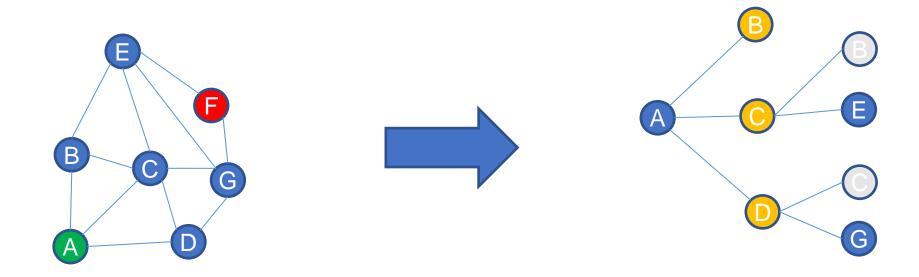
Update cost for **all edges up to current depth** before proceeding to greater depth

- Complete (will find the solution if it exists)
- Guaranteed to find the shortest (number of edges) path
- First solution found is the optimal path
- Can deal with negative edge (transition) costs

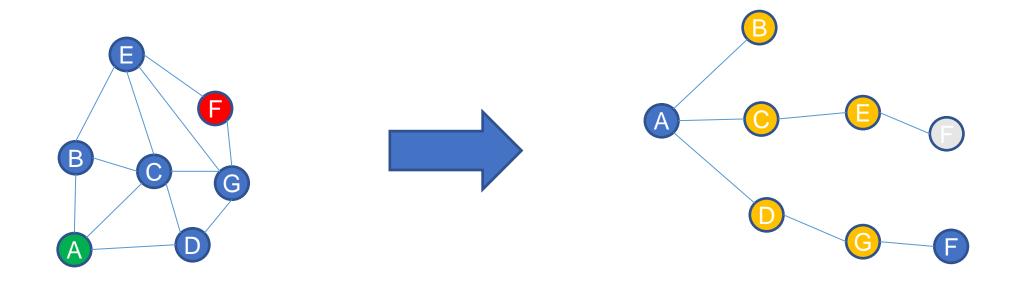
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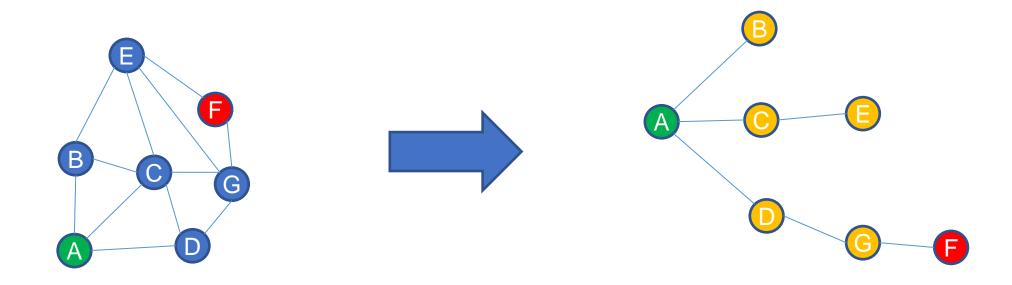
### How to get next node?



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#### How to get next node?

Depth-First-Search (DFS): Maintain Q as a **stack** – Last in/first out

Lower memory requirement (only need to store part of graph)

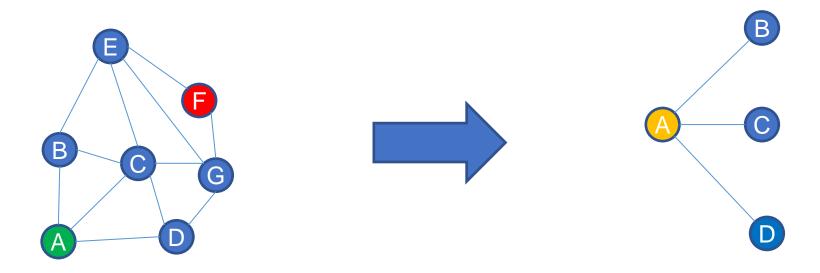
Starts at the root node and explores as far as possible along each branch before backtracking

- Lower memory footprint than BFS with high-branching
- Both BFS and DFS are simple to implement (but generally inefficient).
- not complete for infinite trees

c: frontier/open/alive set of nodes = next nodes to be visited

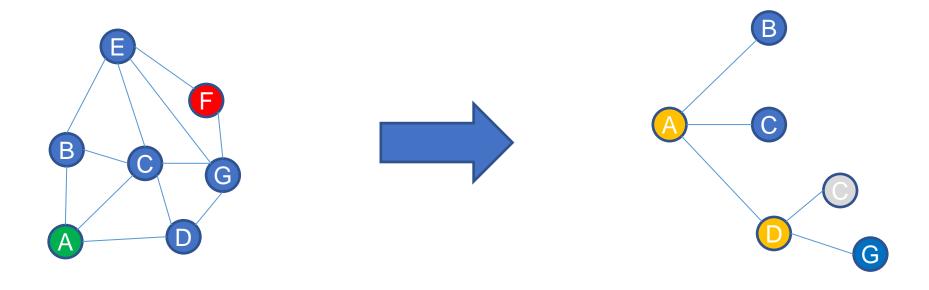
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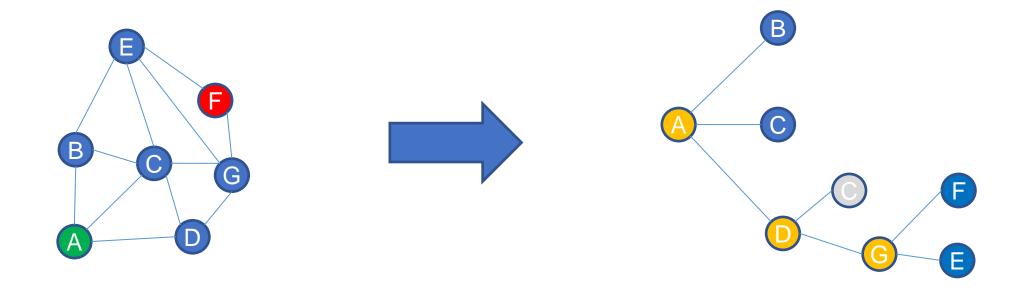
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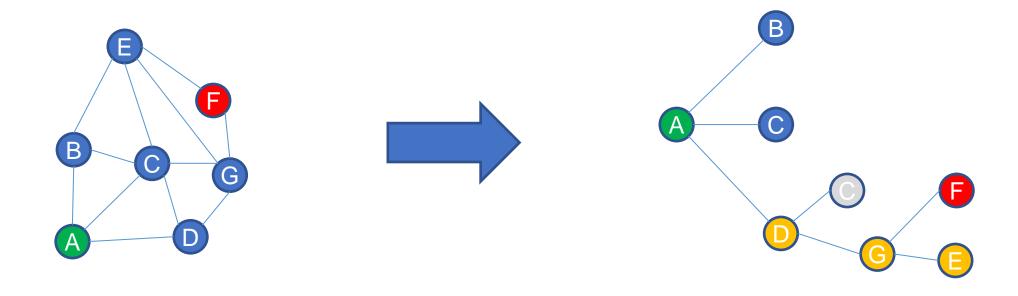
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Depth-First-Search (DFS): Maintain Q as a **stack** – Last in/first out



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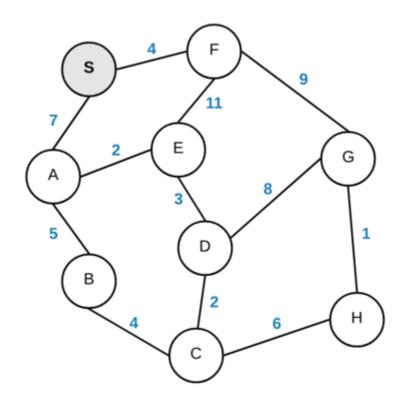
### Dijkstra

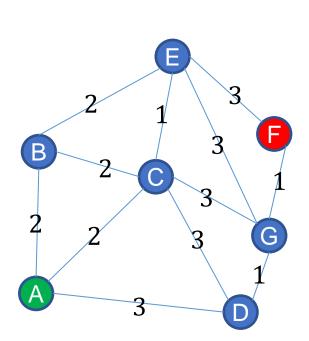
Best-First (BF, Dijkstra): next  $q: \rightarrow q = argmin_{q \in Q} C(q)$ 

- Node will enter Q at most once
- Requires costs to be non-negative

Expanding from closest to start (BFS with edge costs)

Q is ordered according to <u>currently known</u> best cost to arrive



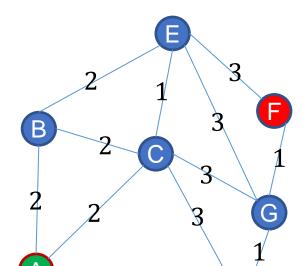


How it works

Q Checked

A(0)

Add in Q ordered by cost

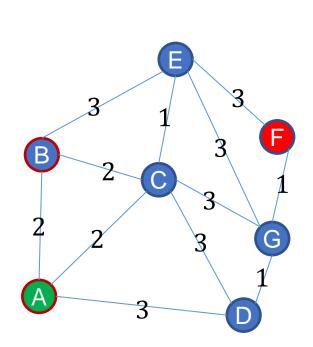


How it works

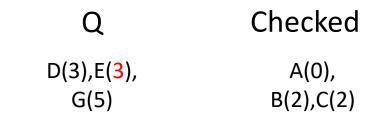
Move in checked; place connected in Q ranked by cost

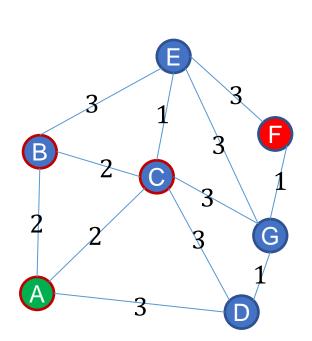
#### How it works



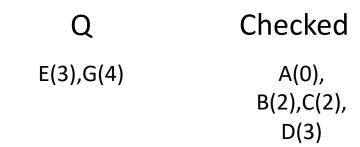


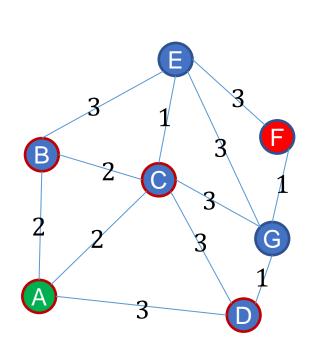
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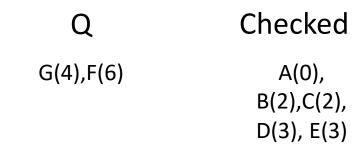


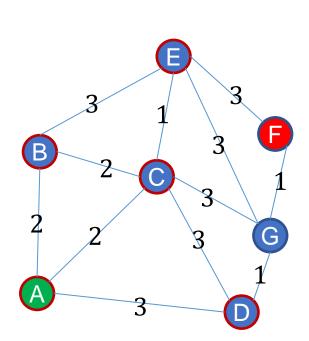
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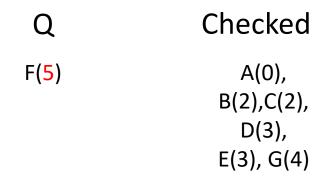


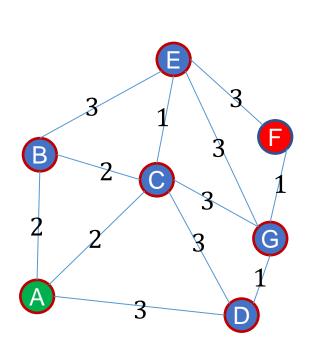
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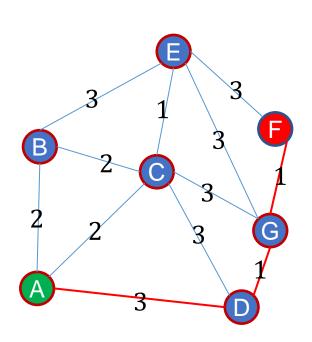


#### How it works





### How it works



Q Checked

[]

A(0),
B(2),C(2),
D(3), E(3),
G(4), F(5) Goal reached

Consider parenting:  $F \leftarrow G \leftarrow A$ 

# **Software requirements**



#### Create Image

- Dimension: ~= 30x30 pixels;
- black walls; grey unknown;
- Save as \*.png (24 bit);

✓ Win: Paint

✓ Ubuntu: KolourPaint

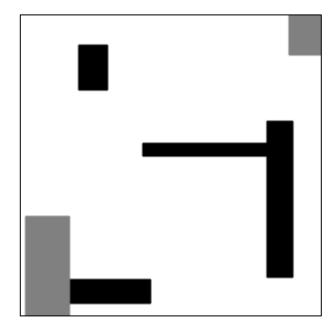
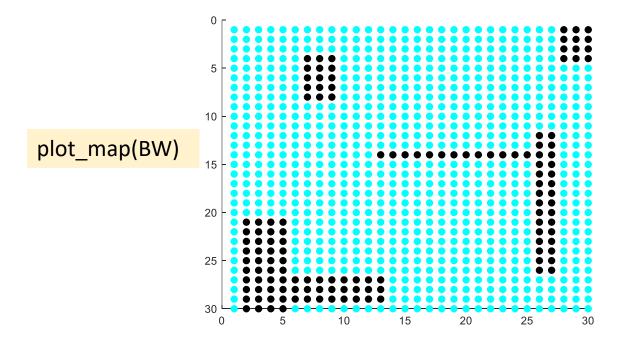


Image-Map

```
map_rgb = imread('mappa_test_red.png');
BW = im2bw(map_rgb,0.7);
```

RGB to bool Matrix (try different threshold values..)



#### Initialize variables & edge - matrix

```
% initialize variables

G=-1*ones(size(BW,1)*size(BW,2));

dist=inf*ones(size(BW,1)*size(BW,2),1);

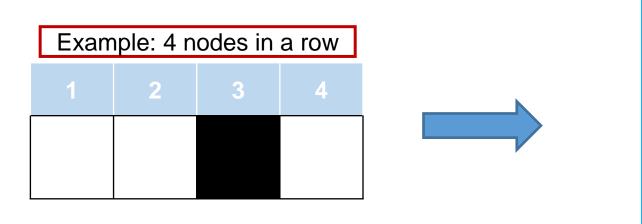
prec=inf*ones(size(BW,1)*size(BW,2),1);

parenting

nodelist=-1*ones(size(BW,1)*size(BW,2),1);

Q (open queue)
```

Edge - matrix



|                 |   | 1  | 2  | 3  | 4  |
|-----------------|---|----|----|----|----|
| Node considered | 1 | 0  | 1  | -1 | -1 |
|                 | 2 | 1  | 0  | -1 | -1 |
|                 | 3 | -1 | 1  | -1 | 1  |
|                 | 4 | -1 | -1 | -1 | 0  |

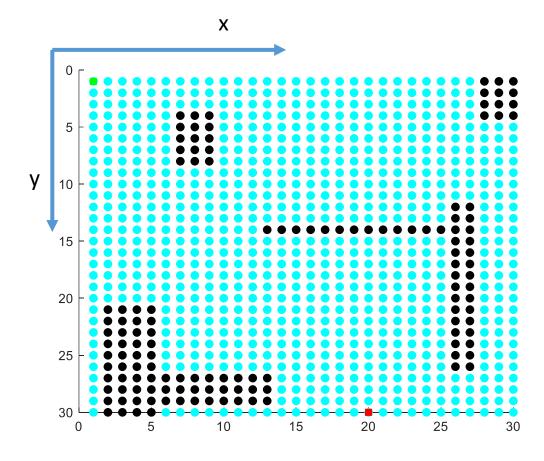
Is it connected to that node?

### Start- goal

```
start_pos = [1,1];
goal_pos = [20,30];

start= (start_pos(2)-1)*size(BW,1)+start_pos(1);
goal= (goal_pos(2)-1)*size(BW,1)+goal_pos(1);

plot(start_pos(1),start_pos(2),'sg','MarkerFaceColor','g')
plot(goal_pos(1),goal_pos(2),'sr','MarkerFaceColor','r')
```



Dijkstra: initialization

Dijkstra: steps 1/2

Q ~= 0 & q ~= goal

evaluate C(q')

update parenting

Dijkstra: steps 2/2

```
[min val,^{\sim}]=min(dist(nodelist(:,1)==1));
    new_nodes=find(dist==min_val);
    tmp i=1;
    while nodelist(new_nodes(tmp_i),1)~=1
         tmp i=tmp i+1;
    end
    act_node=new_nodes(tmp_i);
    nodelist(act_node,1)=0;
    [~,con_nodes]=find(G(act_node,:)>0);
    i con=length(con nodes);
    while i con>0
         if nodelist(con_nodes(i_con),1) ~= 0
                   nodelist(con nodes(i con),1) = 1;
         end
         i_con=i_con-1;
    end
end
```

select q' in Q according to C(q')

remove q from Q

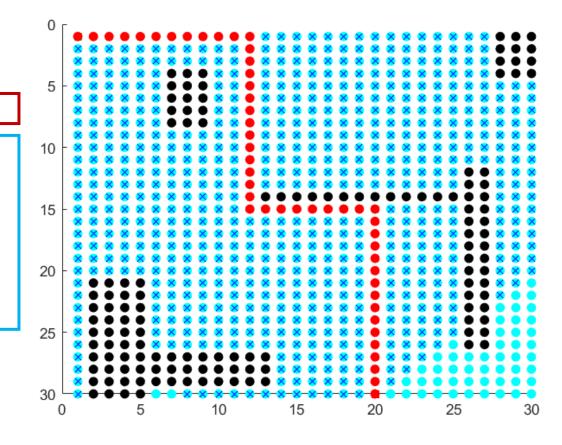
Find connected & add in Q

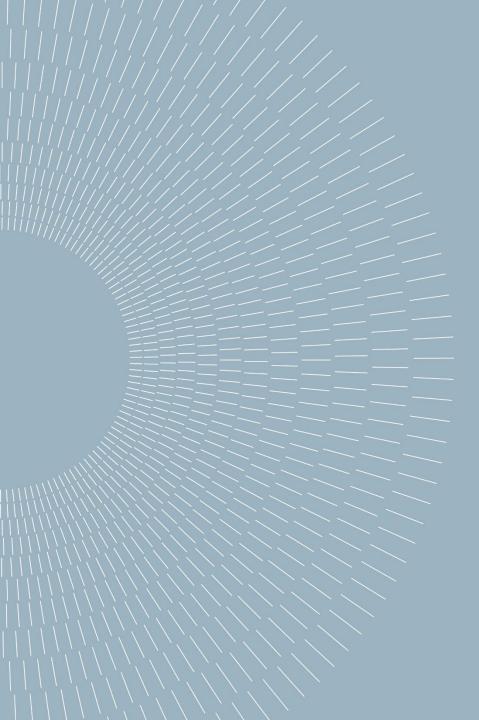
Dijkstra: shortest path

```
if dist(goal)<inf
     sol_id=goal;
    path=[];
    while(sol_id~=start)
        path=[sol_id path];
        sol_id=prec(sol_id);
    end
    path=[sol_id path];</pre>
```

If a solution is available

recorsively define the sequence







**Assignment IV** 

# What we expect from you

# Starting from initial Dijkstra code (v-I) provided:

- 1. Modify the code to consider diagonals movements as allowed (v-II);
- 2. Modify the code to implement A\* algorithm (v-III);
- 3. Implement both (v-IV).

Apply the algorithms (all) to the maps created from images (estimate also start&goal) provided and compare the results.

# Hints: diagonals & A\*

#### **Diagonals:**

- Adjacent nodes dist = 1;
- Diagonal nodes dist =  $\sqrt{2}$ ;

#### <u>A\*:</u>

```
Dijkstra orders by optimal "cost-to-arrival" A* orders by "cost-to-arrival"+ (approximate) "cost-to-go" C(q') = C(q) + C(q,q') + h(q') where h(q') is an underestimate of "cost-to-go
```

#### Results

- O What is mandatory for the report?
  - Plot of shortest path;
  - Evaluate total number of nodes analyzed;
  - Plot the sequence of «actual nodes» computed
  - (compare & comment results)

