

# Introduction of Grid-Based Planning

Grid-Based Planning  
RCI Lab



Speaker: Taehyun Jung

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- | 2. Occupancy Grid
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01

# What is the Grid?

# 01

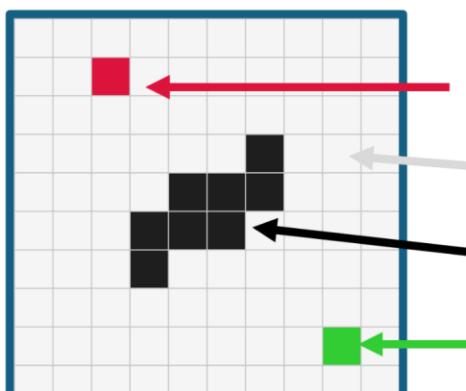
## What is the Grid?

### Grid

A model that simplifies a complex continuous space into a set of **uniform cells** that are easy for a computer to understand.

### Key Features of Grid

- **Discrete Space:** Represents the world as a set of **distinct cells**.
- **Simplicity:** Each cell holds **simple information**, such as ‘traversable’ or ‘obstacle’.
- **Coordinates:** Every cell has a **unique address**, like  $(x, y)$ .



< Grid Map >

Goal point  
Traversable  
Obstacle  
Start point

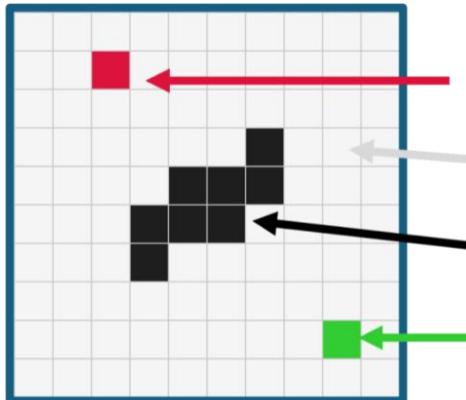
Finite State Space

Make it easy to compute

## 01

# What is the Grid?

- Simplicity:** Each cell holds **simple information**, such as traversable or obstacle.
- Coordinates:** Every cell has a **unique address**, like (x, y).



< Grid Map >

Goal point

Traversable

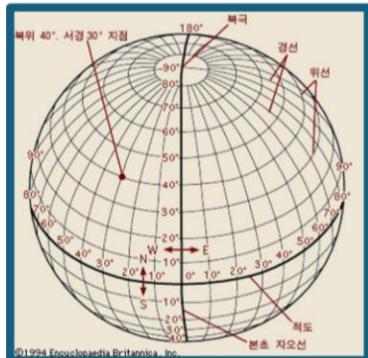
Obstacle

Start point



Finite  
State Space

Make it easy  
to compute



< Finite State Space of Earth >

I'm located at 37.5665 degrees north latitude and 126.9780 degrees east longitude.



## 01

# What is the Grid?

How many meters per cell?

## Resolution ( $r$ )

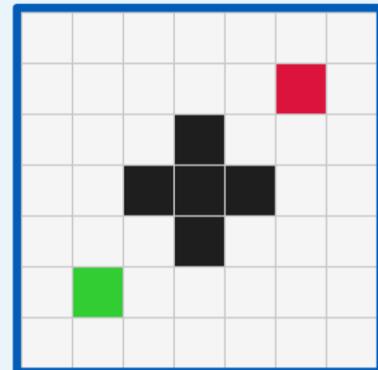
- Definition: how much real-world distance a single grid cell represents.
- Unit: meters per cell [m/cell] ※ There can be another length scale(or unit) instead of meters
- World  $(x, y) \rightarrow$  Grid  $(i, j)$ , total counts of cells  $N$

$$i = \left\lfloor \frac{y - y_{\min}}{r} \right\rfloor, \quad j = \left\lfloor \frac{x - x_{\min}}{r} \right\rfloor, \quad N \approx \frac{L_x L_y}{r^2} \quad \begin{cases} L_x : \text{length of width} \\ L_y : \text{length of height} \end{cases}$$

- It directly controls detail, computation, and memory.

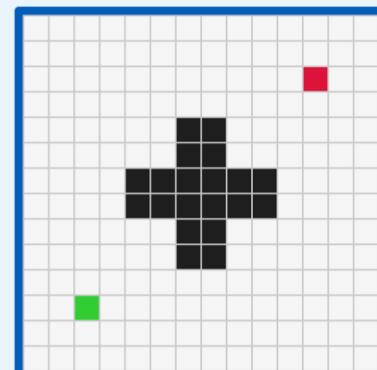
### < Low Resolution >

- Fewer cells
- Detail ↓**
- Computation ↓**
- Memory ↓**



### < High Resolution >

- More cells
- Detail ↑**
- Computation ↑**
- Memory ↑**



## 01

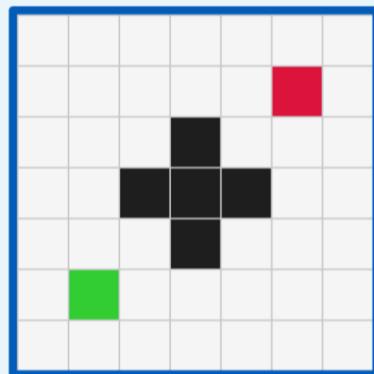
# What is the Grid?

How many meters per cell?

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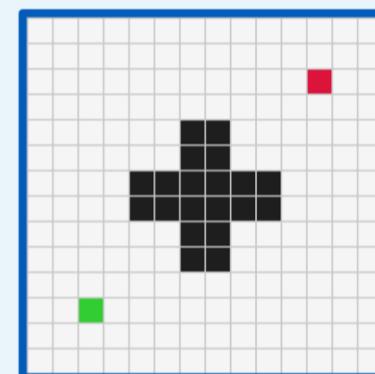
< Low Resolution >

- Fewer cells
- Detail ↓
- Computation ↓
- Memory ↓



< High Resolution >

- More cells
- Detail ↑
- Computation ↑
- Memory ↑



Example)

- Map: Width( $L_x$ ) = 1m, Height( $L_y$ ): 1m
- Resolution:  $r_{\text{low}} = 1/7$  [m/cell],  $r_{\text{high}} = 1/14$  [m/cell]

$$\left. \begin{array}{l} N_{\text{low}} = 49 \text{ cells} \\ N_{\text{high}} = 196 \text{ cells} \end{array} \right\} \rightarrow$$

- The computational time complexity of Dijkstra/A\* in the worst case:  $O(N \log N)$

02

# Occupancy Grid

# 02

# Occupancy Grid

## Occupancy Grid

- A map that encodes whether each grid cell is occupied by an obstacle.

## Representation

State	Integer Convention (ROS)	Probabilistic	Meaning
Free space	0	Low $p_{occ}$	Traversable
Occupied space	100	High $p_{occ}$	Non-traversable (collision)
Unknown space	-1	Undetermined $p_{occ}$	Unobserved or Uncertain area

- For probabilistic maps, choose a threshold  $\tau$ , declare occupied if  $p_{occ} \geq \tau$ .
- We should choose the policy for unknown area :
  1. **Conservative**: treat as blocked or assign very high cost (avoid exploration into unknown).
  2. **Aggressive**: treat as free (faster paths but higher collision risk).
  3. **Compromise**: assign medium-to-high cost and let the planner decide.

# 02

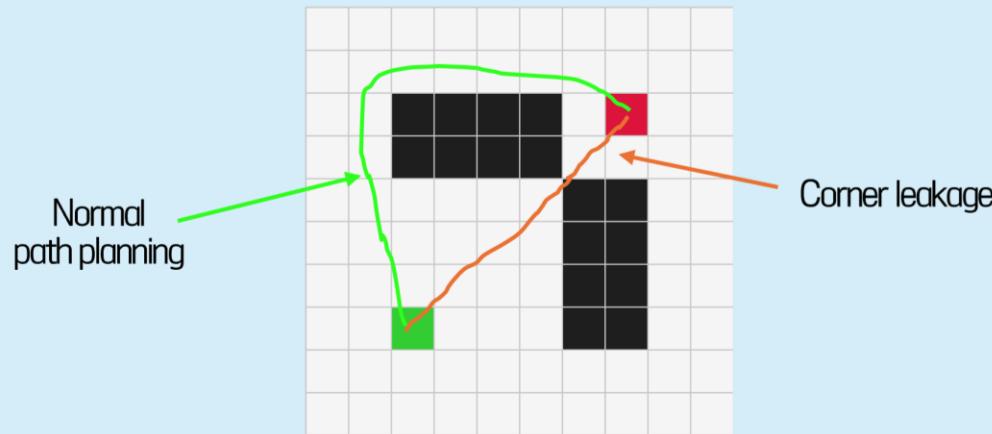
# Occupancy Grid

## Occupancy Grid

- A map that encodes whether each grid cell is occupied by an obstacle.

## Limitations

- Occupancy grid only represent whether robot can pass.
- Corner leakage on diagonals:



→ Solution: Combine occupancy with **connectivity rules** and **inflation**.

03

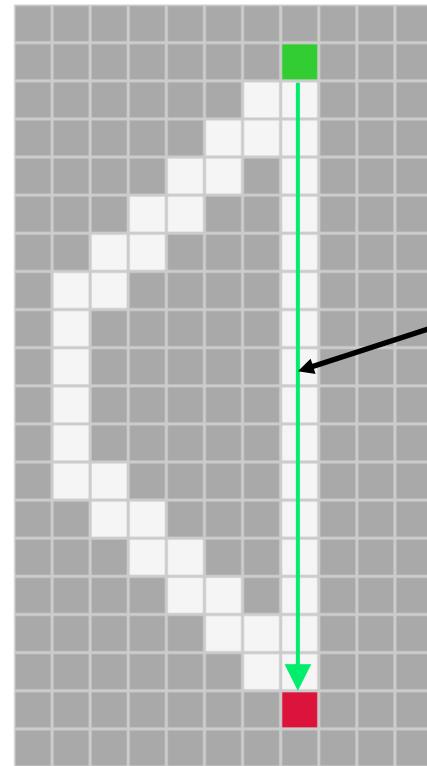
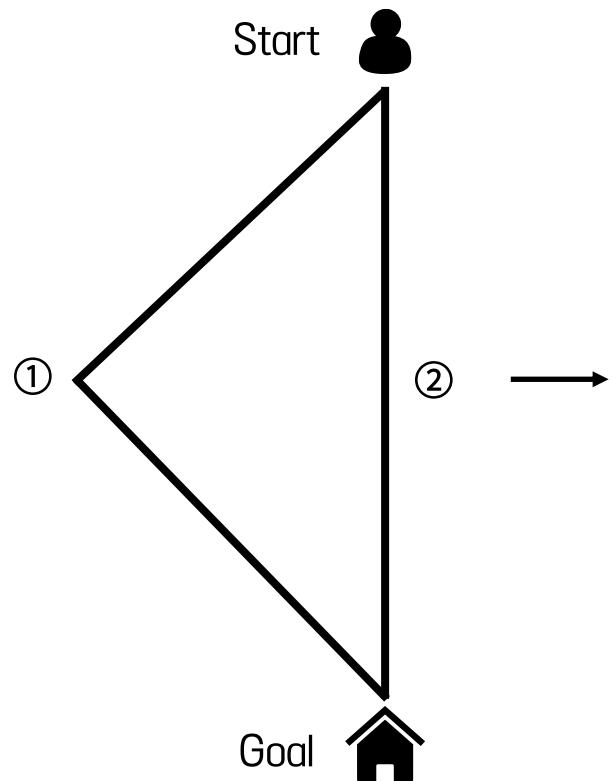
# Costmap

# 03

## Costmap

### Cost

- A scalar score for **undesirability**.
- Smaller is better.**



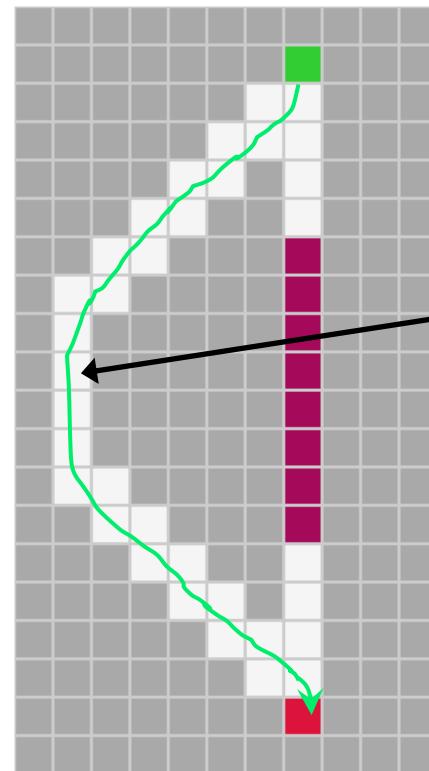
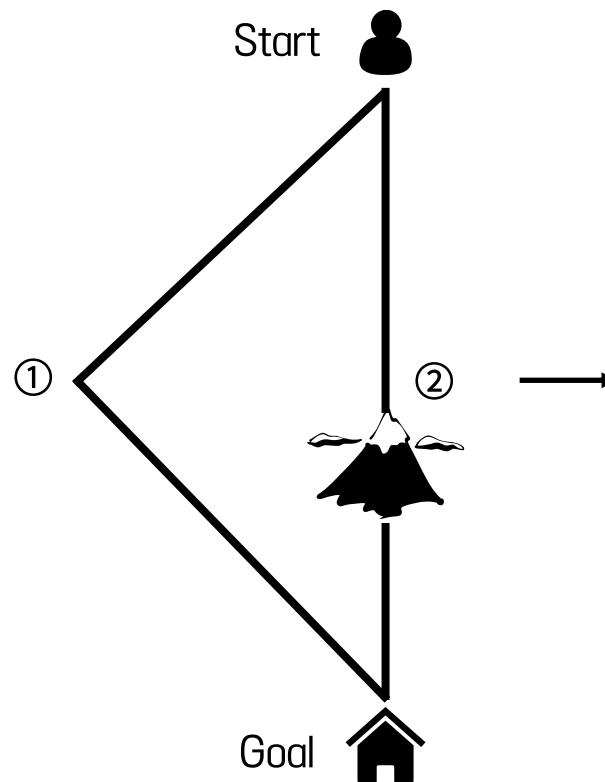
Anyone would choose path ② which is the shortest way.

## 03

## Costmap

## Cost

- A scalar score for **undesirability**.
- Smaller is better.**



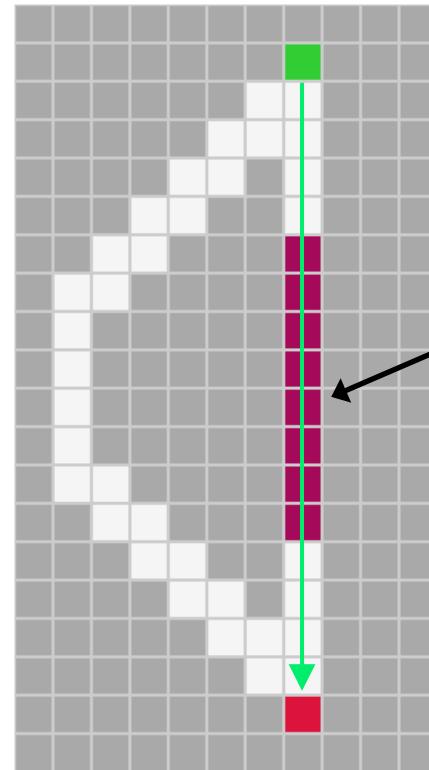
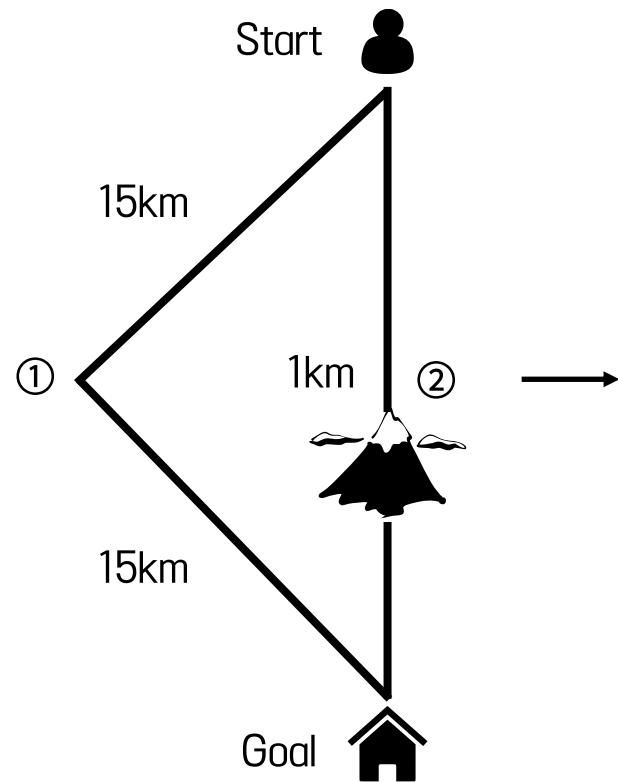
Even if path ① is farther than ②,  
anyone would choose ①  
to avoid high mountains.

# 03

# Costmap

## Cost

- A scalar score for **undesirability**.
- Smaller is better.**



If path ① is too far,  
it may be better to climb the mountains.

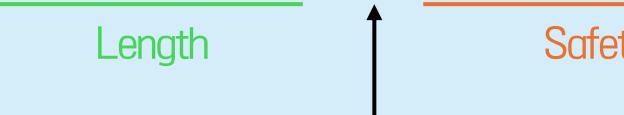
# 03

# Costmap

## Cost

$$\text{PathCost} = \sum_{(u \rightarrow v \in \text{path})} \text{Cost}(u \rightarrow v)$$

where  $\text{Cost}(u \rightarrow v) = \frac{\text{move\_cost}(u, v)}{\text{Length}} + \beta \cdot \frac{\text{sample}(c(u), c(v))}{\text{Safety}}$



Trade-off between **length** and **safety**

## \* Reference of the distance

	Manhattan	Octile	Chebyshev
Definition	$ x_1 - x_2  +  y_1 - y_2 $	$\sqrt{2} \times \min(\Delta x, \Delta y) +  \Delta x - \Delta y $	$\min( \Delta x_1 - \Delta x_2 ,  \Delta y_1 - \Delta y_2 )$
Characteristic	Used when robots can only travel in certain directions	Particularly useful for path planning of diagonal-moving robots in grid based planning	Suitable when assuming diagonal travel costs equal to straight travel costs

# 03

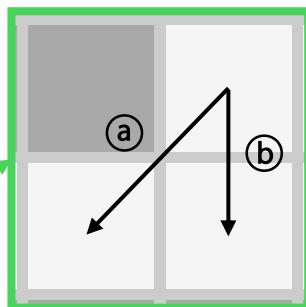
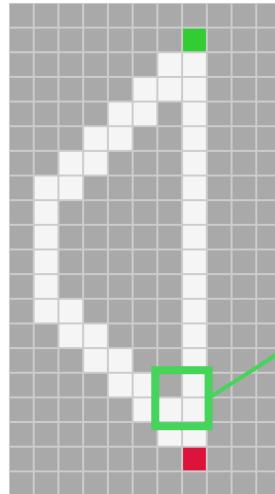
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Trade-off between **length** and **safety**



In this tutorial,  
Manhattan & Octile distance will be used.

ⓐ Diagonal:  $\text{move\_cost}(u, v) = \sqrt{2}$

ⓑ Straight:  $\text{move\_cost}(u, v) = 1$

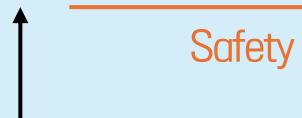
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# Costmap

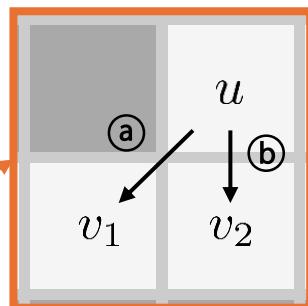
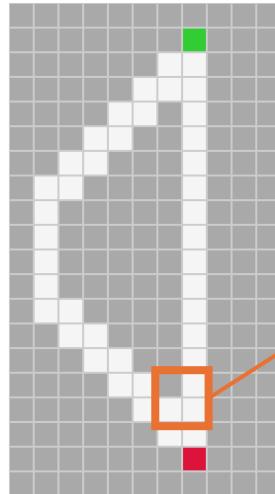
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Trade-off between **length** and **safety**



ⓐ :  $\text{sample}(c(u), c(v_1)) = \frac{c(u) + c(v_1)}{2}$

ⓑ :  $\text{sample}(c(u), c(v_2)) = \frac{c(u) + c(v_2)}{2}$

\* Methods of calculating costs are determined by trade-offs such as speed, safety, and complexity.

# 03

# Costmap

## Cost Convention in ROS

Cost uses an 8-bit integer(0 ~ 255) scale in ROS Navigation stack(ROS1 costmap\_2d, ROS2 Nav2).

Value	Meaning	Note
0	Free space	Traversable
1 ~ 252	Inflation/Potential	The closer to the obstacle, the larger cost. (Detailed values are determined by Inflation layer.)
253	Inscribed inflated obstacle	Substantial collision range. Treat it an evasive/prohibitive level when planning.
254	Lethal obstacle	<b>Actual obstacle core.</b> Treat it a prohibitive level when planning.
255	No information (Unknown)	Unobserved/uncertain areas. According to policy, it is blocked or set high cost.

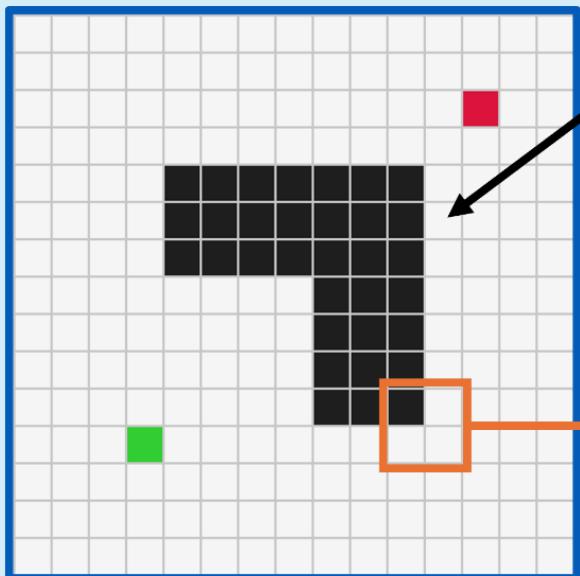
# 03

# Costmap

## Costmap

- A scalar **cost field** built from occupancy (and optionally other layers).
- Occupancy grid tells only **can/can't pass**.  
Costmap adds **how preferable** each traversable cell is, combining **shortness vs. safety**.

## Components of costmap



Obstacle core

Occupied cells are **non-traversable** and should receive a **very large** cost  $C_{obs}$  in your chosen scale.

254	0
0	0

ex) 255 scale: fixed to 254

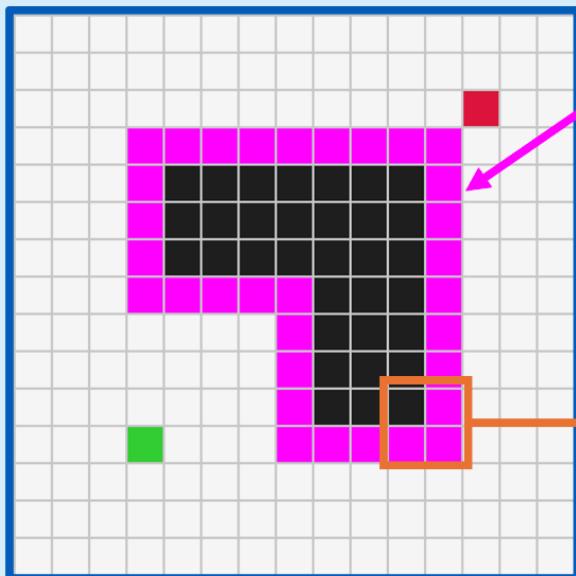
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## Components of costmap



### Inflation ring

Around each obstacle, assign a distance-dependent penalty.  
The closer a free cell is to an obstacle, the higher its cost.  
This **prevents corner-cutting paths** that graze obstacles.

254	200
200	180

# 03

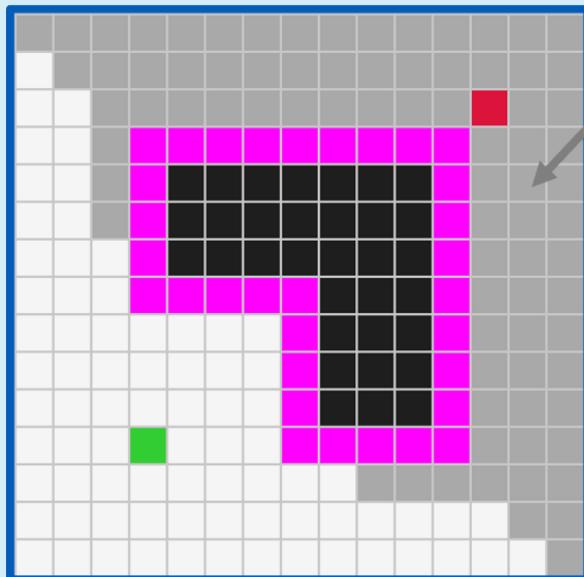
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A scalar cost field built from occupancy (and optionally other layers).

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## Components of costmap



Unknown

If you decided not to hard-block Unknown in the occupancy layer, you can encode **caution here** by assigning a high but finite cost to unknown cells.

This keeps exploration **possible** yet **discouraged** compared to well-observed free space.

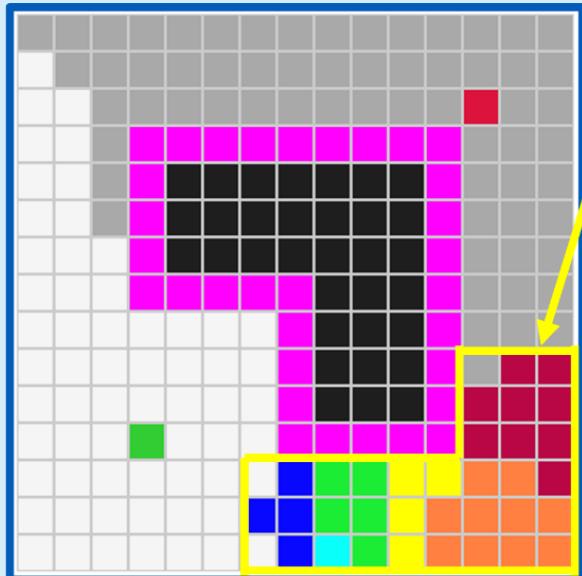
# 03

# Costmap

## Costmap

- ▶ A scalar cost field built from occupancy (and optionally other layers).
- ▶ Occupancy grid tells only **can/can't pass**.  
Costmap adds **how preferable** each traversable cell is, combining **shortness vs. safety**.

## Components of costmap



### Additional layers

Real systems may add layers for slope, sound, radiation, etc., then combine layer via a weighted sum.

For this tutorial, stick to a clean baseline:  
obstacle core + inflation (+ optional unknown)

04

# Inflation

# 04

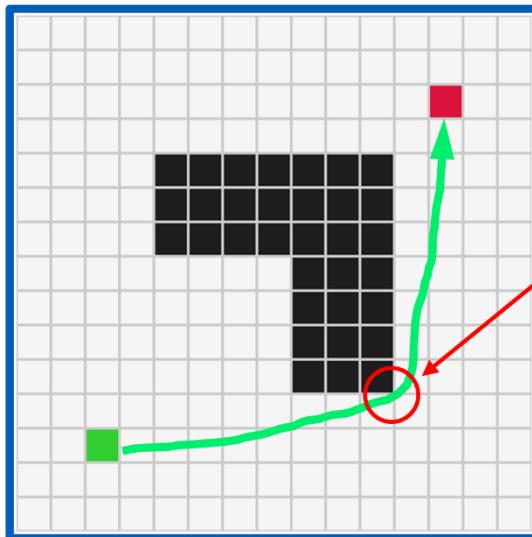
## Inflation

### Why do we need inflation?

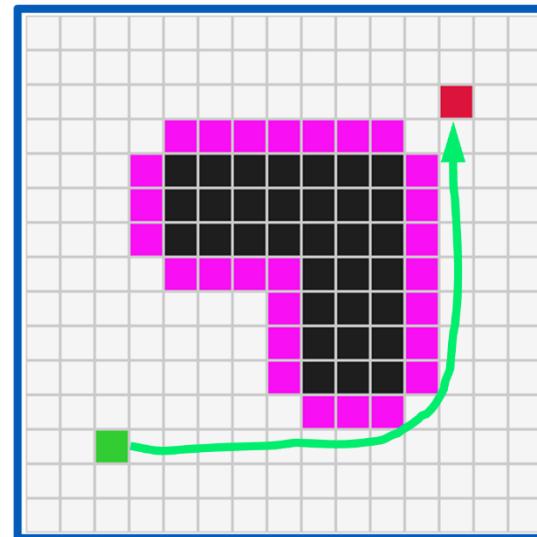
Occupancy grid tells only **can/can't pass**.

Real routes must also consider how far robots keep from obstacles.

Inflation creates a high-cost halo around obstacles so that a planner naturally **maintains clearance** and **reduces unrealistic corner-cutting** using cost only.



< without inflation >



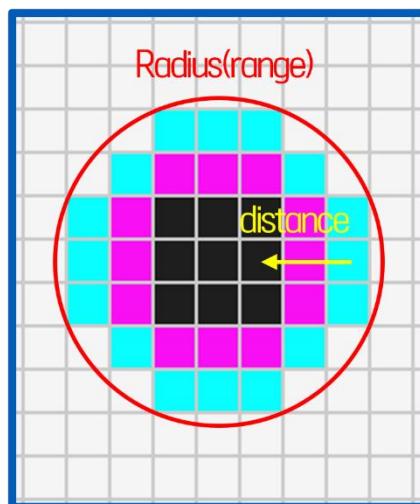
< inflation applied >

# 04

# Inflation

## Core Intuitions of Inflation

- Distance-based cost  
→ Each cell's cost depends on its nearest-obstacle distance  $d$ .
- Closer → Higher cost; Farther → Lower cost  
( $d = 0$  marks the non-traversable core.)
- Radius  $R$  : the effective range of inflation. —> Choose from robot size + margin.  
→ For  $d > R$ , set cost to zero(or very small) so far-away cells remain free.

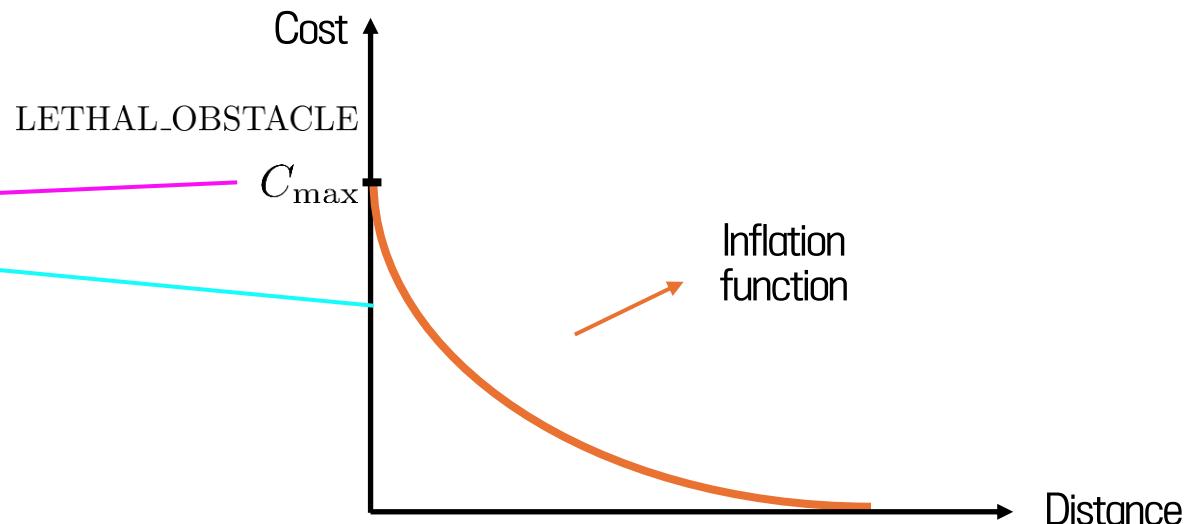
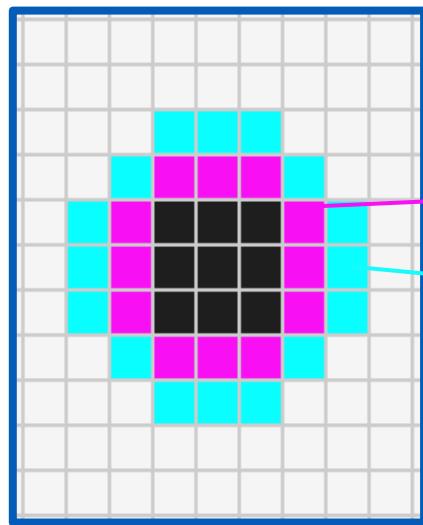


# 04

# Inflation

## Mathematical Models for Inflation Costs

- Design the inflation function to match your intent. → Not a rule!
- Shape the inflation curve to encode your desired clearance and detour preference.
- The essentials are higher cost near obstacles, lower cost farther away. → depending on  $d$

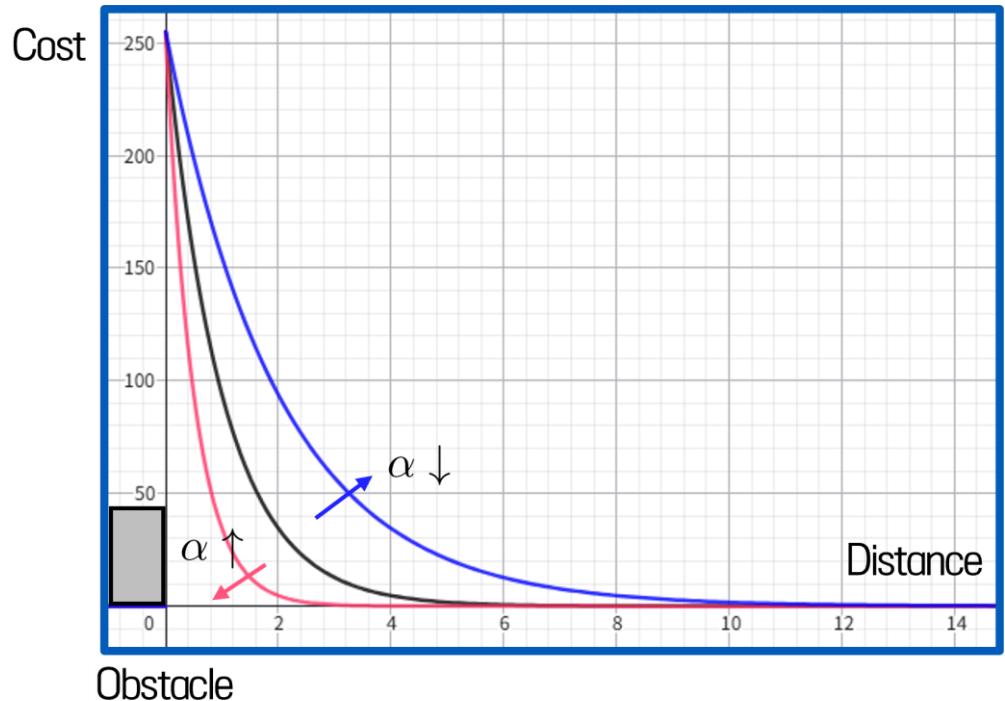


## 04

## Inflation

## Example Models – Exponential Decay

$$C_{\text{infl}}(d) = \begin{cases} \text{Obstacle core} & d = 0 \\ C_{\max}e^{-\alpha d} & 0 < d \leq R \\ 0 & d > R \end{cases}$$



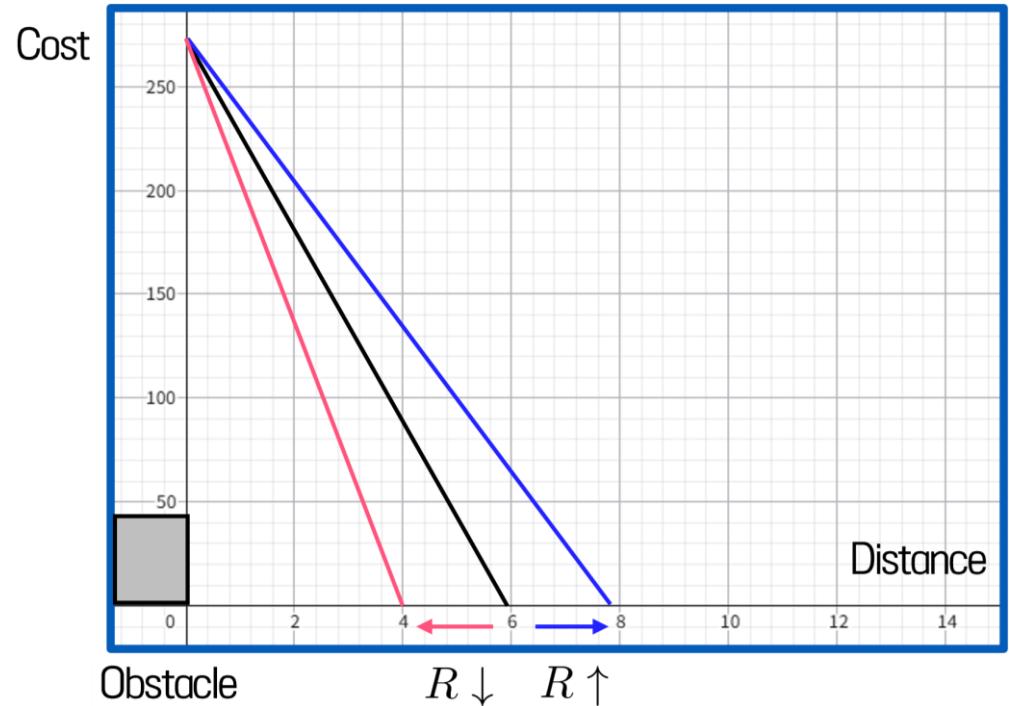
- Strong near the core → drops quickly just a bit away.
- Tuning Parameter  $\alpha$  : strong near-core penalty, fast falloff.
- Exponential method is mainly used.

## 04

## Inflation

## Example Models – Linear Lamp

$$C_{\text{infl}}(d) = C_{\max} \max(0, 1 - \frac{d}{R})$$



- Linear decrease → very intuitive and predictable.

# 04

# Inflation

## Application of Inflation in a Real ROS/Nav2 Environment

Location: /opt/ros/humble/include/nav2\_costmap\_2d/nav2\_costmap\_2d/inflation\_layer.hpp

```

144
145  /** @brief Given a distance, compute a cost.
146  * @param distance The distance from an obstacle in cells
147  * @return A cost value for the distance */
148 inline unsigned char computeCost(double distance) const
149 {
150     unsigned char cost = 0;
151     if (distance == 0) {
152         cost = LETHAL_OBSTACLE; → 254
153     } else if (distance * resolution_ <= inscribed_radius_) {
154         cost = INSCRIBED_INFLATED_OBSTACLE; → 253
155     } else {
156         // make sure cost falls off by Euclidean distance
157         double factor =
158             exp(-1.0 * cost_scaling_factor_ * (distance * resolution_ - inscribed_radius_));
159         cost = static_cast<unsigned char>((INSCRIBED_INFLATED_OBSTACLE - 1) * factor);
160     }
161     return cost;
162 }
```

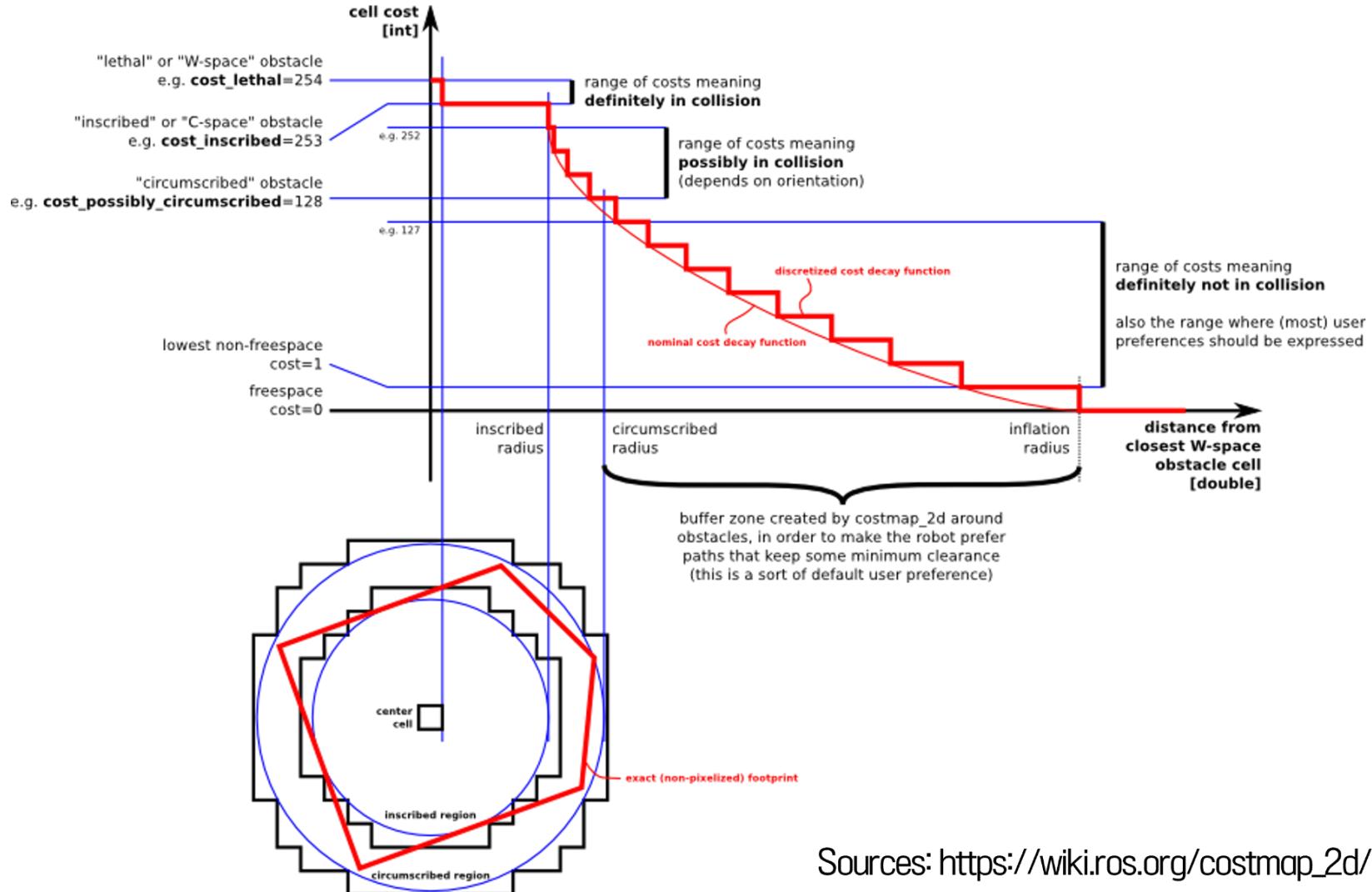
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<Exponential Decay>

# 04

# Inflation

## Application of Inflation in a Real ROS/Nav2 Environment



05

# Connectivity

# 05

# Connectivity

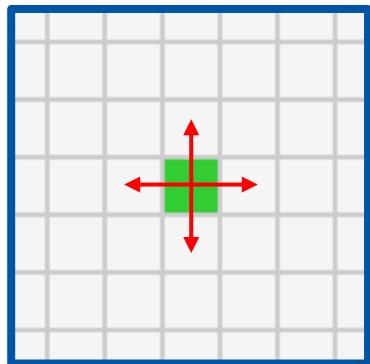
What does connectivity mean in path planning?

Connectivity defines which neighbor cells are allowed to move to from a given cell.

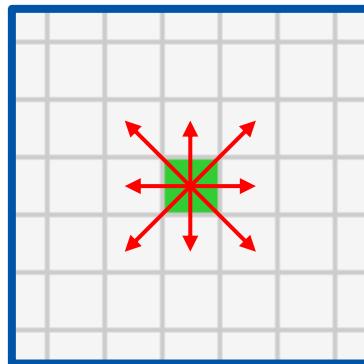
Standard Choices of Connectivity

- 4-connected (Von Neumann) : Up, Down, Left, Right → Manhattan
- 8-connected (Moore) : 4-connected + 4 diagonals → Octile

\* With 8-connected moves, paths can be shorter and straighter. However, it can be more expensive of more branching.



<4-connected>



<8-connected>

$$C_{\text{step}} = \begin{cases} 1 & \text{cardinal moves} \\ \sqrt{2} & \text{diagonals (if 8-connected)} \end{cases}$$

<per-step cost>

# 05

# Connectivity

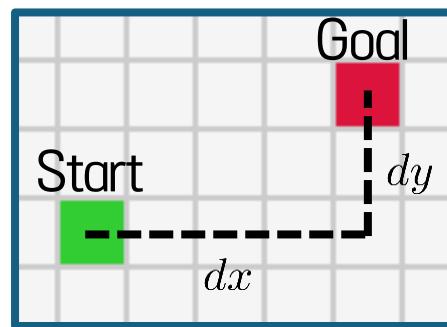
## Heuristic Alignment (A\* Preview)

- 4-connected → Manhattan Distance (admissible/consistent)

$$h = |dx| + |dy|$$

- 8-connected → Octile Distance (with straight = 1, diagonal =  $\sqrt{2}$ )

$$h = (|dx| + |dy|) + (\sqrt{2} - 2) \cdot \min(|dx|, |dy|)$$



# 05

# Connectivity

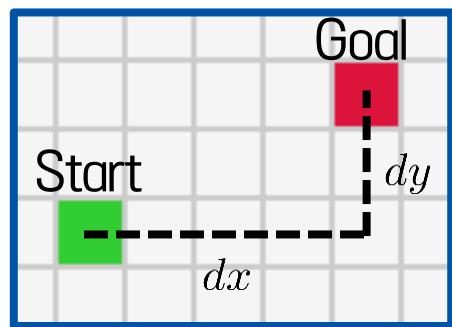
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$$\begin{aligned} dx &= 4 \\ dy &= 2 \end{aligned}$$



< 4-connected >

$$\begin{aligned} h &= |dx| + |dy| \\ &= |4| + |2| \\ &= 6 \end{aligned}$$

# 05

# Connectivity

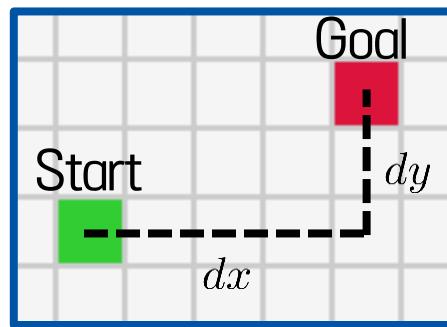
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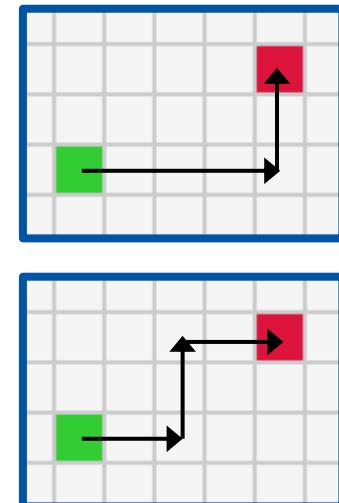
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< 4-connected >

$$\begin{aligned} h &= |dx| + |dy| \\ &= |4| + |2| \\ &= 6 \end{aligned}$$



⋮

# 05

# Connectivity

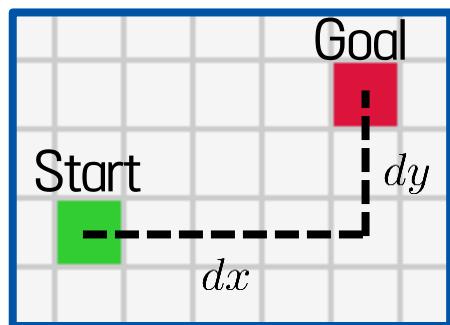
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$$h = (|dx| + |dy|) + (\sqrt{2} - 2) \cdot \min(|dx|, |dy|)$$



< 8-connected >

$$\begin{aligned}
 h &= (|dx| + |dy|) + (\sqrt{2} - 2) \cdot \min(|dx|, |dy|) \\
 &= (|4| + |2|) + (\sqrt{2} - 2) \cdot |2| \\
 &= 2 + 2\sqrt{2}
 \end{aligned}$$

# 05

# Connectivity

## Heuristic Alignment (A\* Preview)

- 4-connected → Manhattan Distance (admissible/consistent)

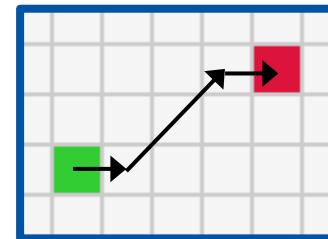
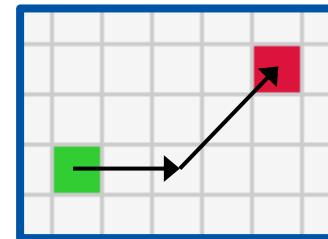
$$h = |dx| + |dy|$$

- 8-connected → Octile Distance (with straight = 1, diagonal =  $\sqrt{2}$ )

$$h = (|dx| + |dy|) + (\sqrt{2} - 2) \cdot \min(|dx|, |dy|)$$

<8-connected>

$$\begin{aligned} h &= (|dx| + |dy|) + (\sqrt{2} - 2) \cdot \min(|dx|, |dy|) \\ &= (|4| + |2|) + (\sqrt{2} - 2) \cdot |2| \\ &= 2 + 2\sqrt{2} \end{aligned}$$



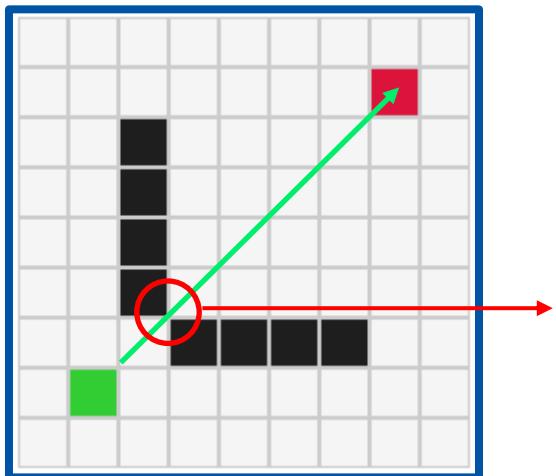
⋮

# 05

# Connectivity

## Diagonal Safety Rule (Anti Corner-Cutting)

Issue: With 8-connected grids, a path may squeeze between two obstacle corners via a diagonal.



There is no problem theoretically, but in real world, it is wrong path.

### Solution(Rule) :

→ Allow a diagonal  $(i, j) \rightarrow (i \pm 1, j \pm 1)$  only if both adjacent cardinal cells  $(i, j \pm 1)$ ,  $(i \pm 1, j)$  are free.

### Effects :

- It robustly prevents corner-cutting.
- It works best together with inflation.

# 05

# Connectivity

## Comparison

### ► 4-connected vs. 8-connected

	4-connected	8-connected
Branching	Lower (often faster)	Higher (more list expansions, computation ↑)
Path Length & Straightness	Can be longer	Often shorter & straighter
Corner-cut Risk	Low	High (mitigate with safety rule)
Heuristic	Manhattan	Octile
Implementation	Simple	Moderate (add safety rule)