

# UAV Formation Base Path Planning

Presented by Group 1

# TEAM MEMBERS

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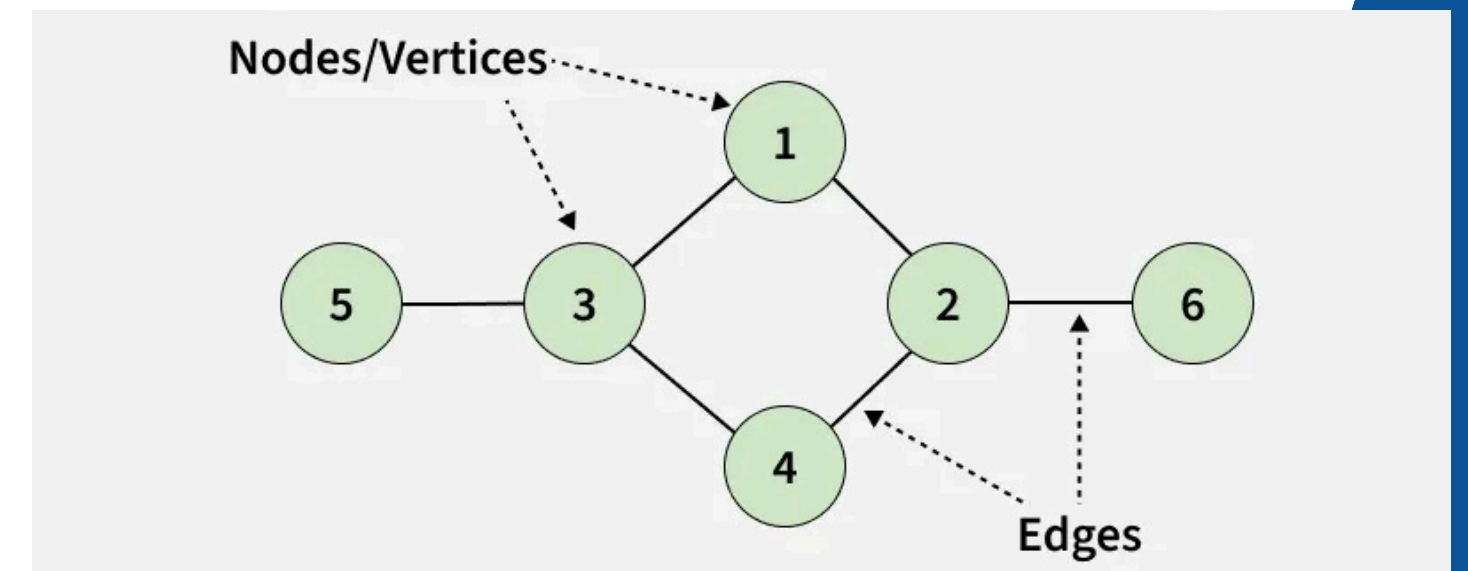
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# WHAT IS A GRAPH?

- A graph models relationships between entities
- Two main components:
  - a. Vertices (nodes): entities or states
  - b. Edges: connections or transitions
- Graphs appear everywhere:
  - a. Road networks
  - b. Social networks
  - c. Computer networks
  - d. State spaces in AI and robotics

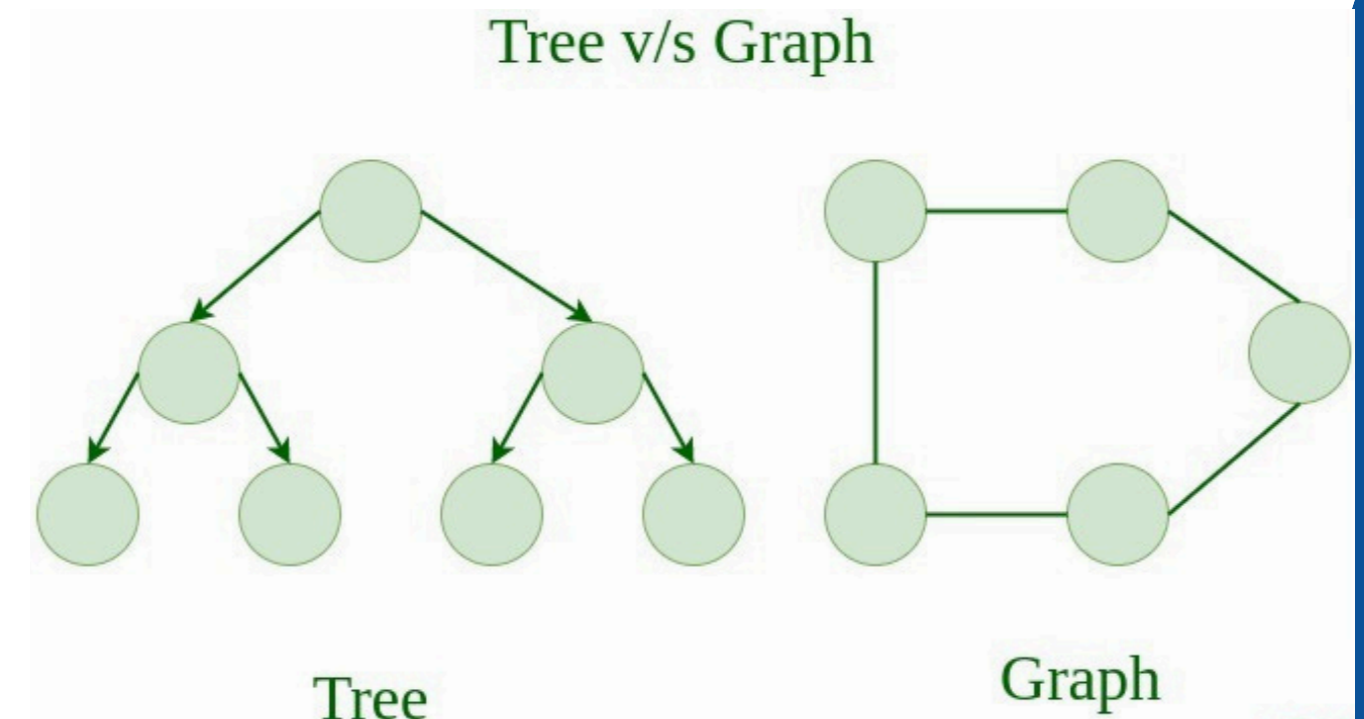


# TYPES OF GRAPHS

- Directed vs Undirected
  - Undirected: edges work both ways (roads)
  - Directed: edges have direction (one-way streets, followers)
- Unweighted vs Weighted
  - Unweighted: all edges equal cost
  - Weighted: edges have costs (distance, time, energy)
- Sparse vs Dense (idea level)
  - Sparse: few edges per node
  - Dense: many edges per node

# TREES VS GENERAL GRAPHS

- Tree:
  - No cycles
  - Exactly one unique path between any two nodes
  - $N$  nodes  $\rightarrow N-1$  edges
- General graph:
  - May contain cycles
  - Multiple paths between nodes

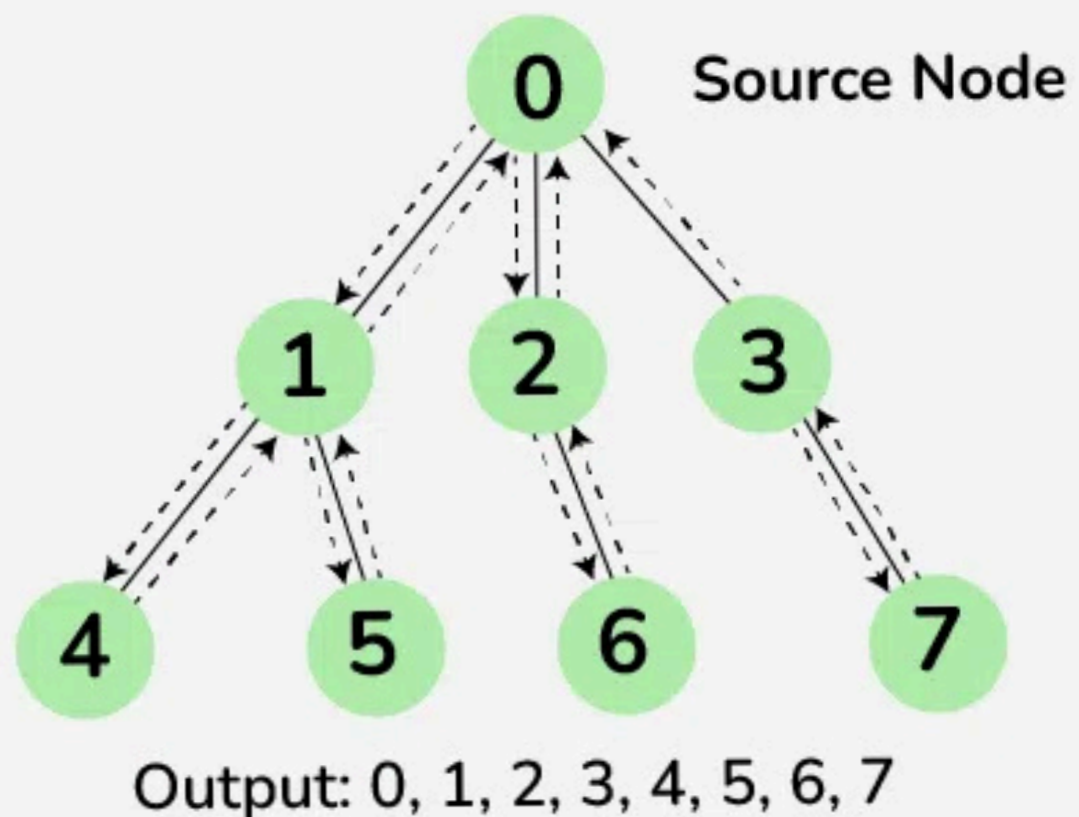


Why this matters:

1. Pathfinding is harder when many paths exist
2. Need systematic traversal algorithms

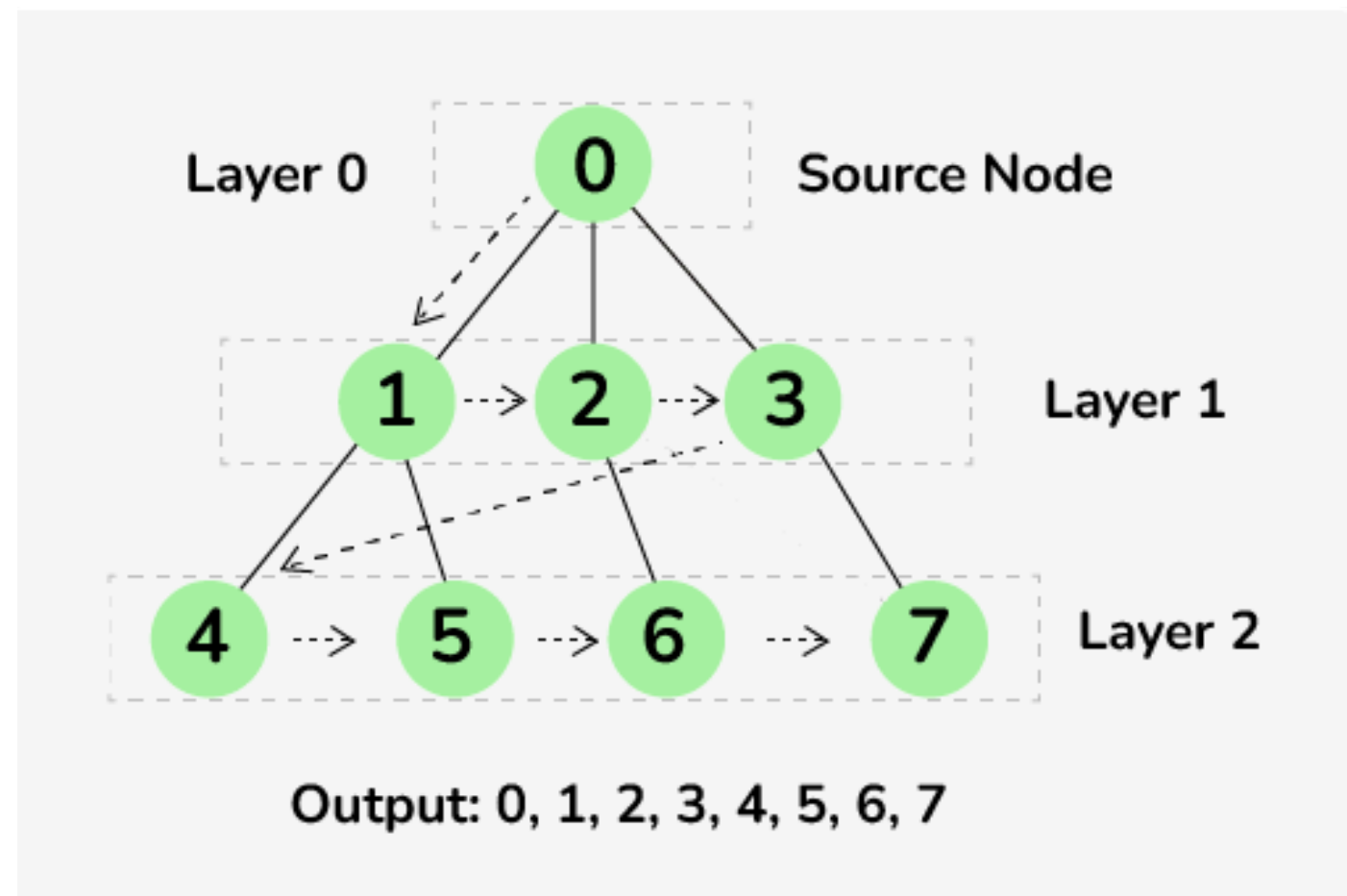


# DEPTH FIRST SEARCH (DFS)



- Explores as far as possible along each branch before backtracking.
- Uses a stack (or recursion).

# BREADTH FIRST SEARCH (BFS)



- Explores all neighbors of a vertex before moving to the next level.
- Uses a queue.

# BFS VS DFS

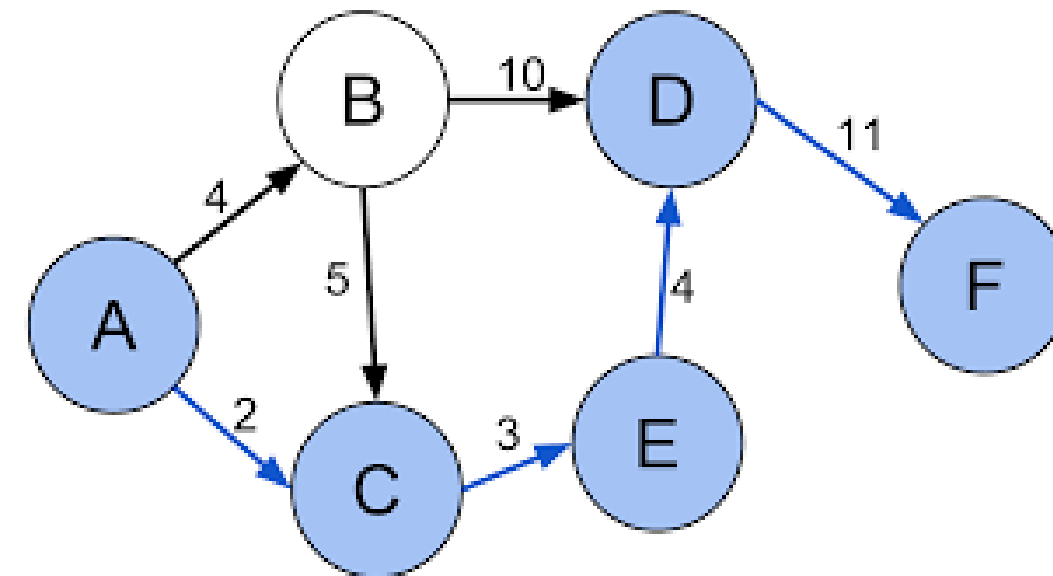
- DFS:
  - Deep, narrow exploration
  - Memory efficient
  - Not optimal for paths
- BFS:
  - Wide, level-based exploration
  - More memory
  - Optimal only for unweighted graphs

Key insight: Real-world problems often have unequal costs.



# NEED FOR WEIGHTED SHORTEST PATH ALGORITHMS

- In many problems, edges have different costs:
  - Distance
  - Time
  - Energy
- BFS treats all edges equally → incorrect results
- We need an algorithm that:
  - Handles non-uniform weights
  - Guarantees optimal paths



This leads to **Dijkstra's Algorithm**

# DIJKSTRA'S ALGORITHM

1. Solves shortest paths in graphs with non-negative weights
2. Core idea:
  - Always expand the node with smallest known distance
3. Uses:
  - Priority queue (min-heap)
4. Guarantees optimality

# A\* ALGORITHM

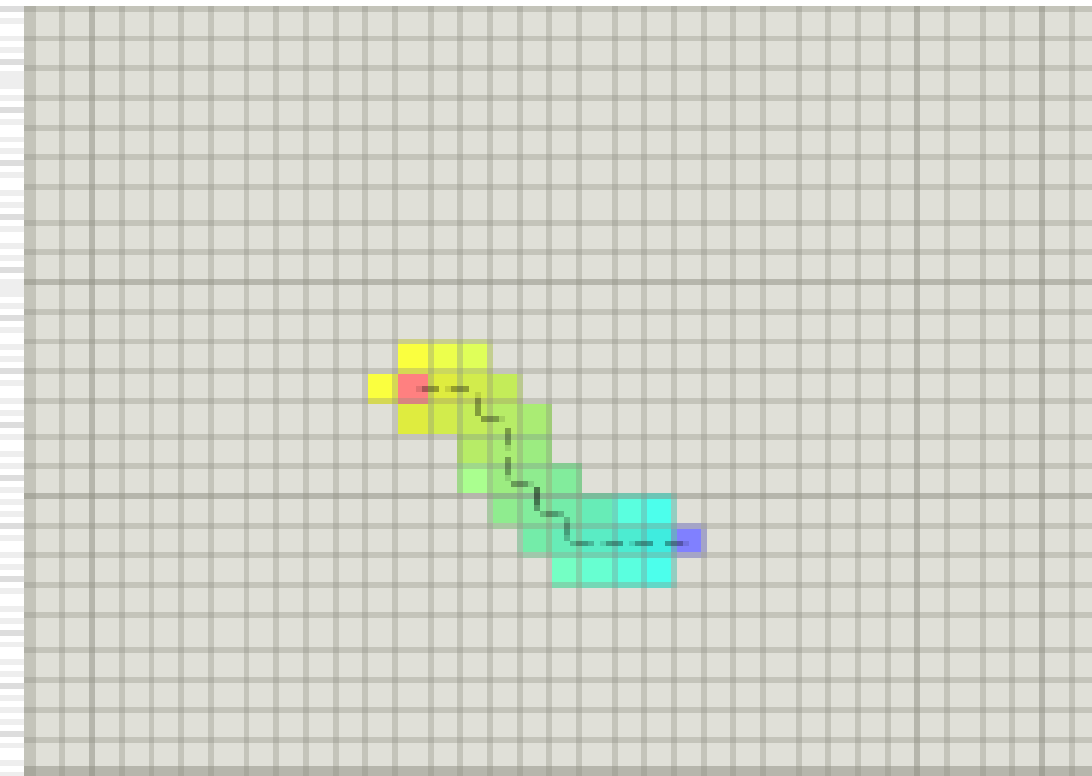
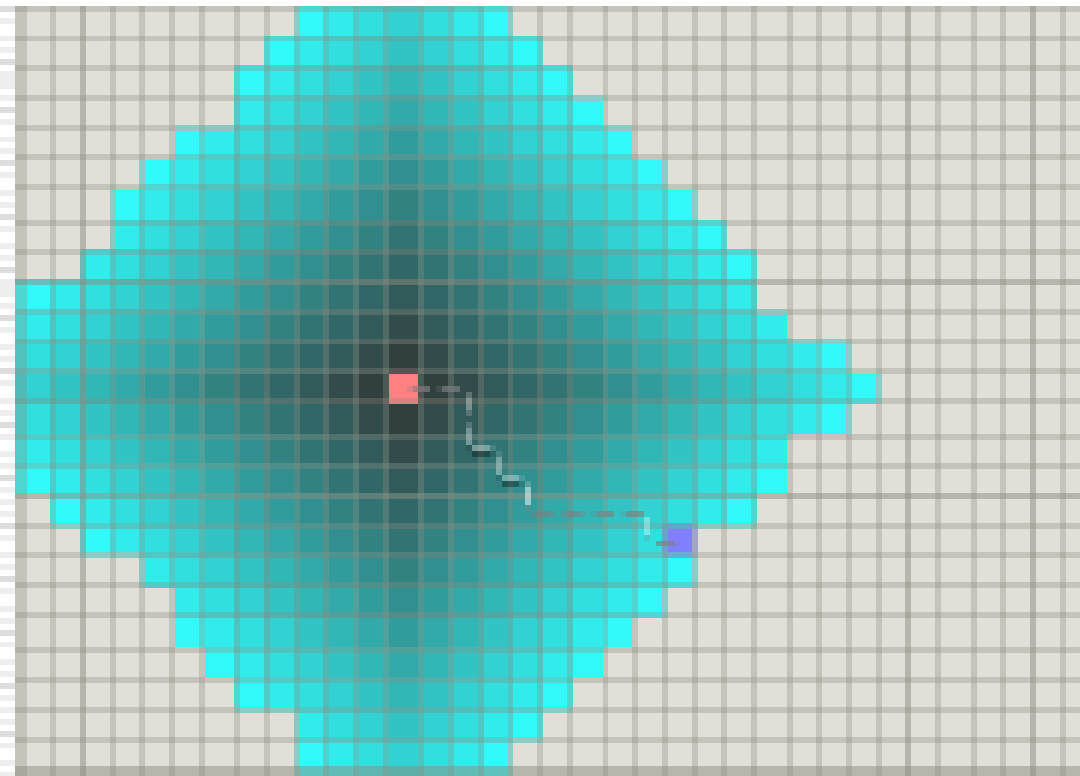
1. Solves shortest paths in graphs using cost + heuristic
2. Core idea:
3. Always expand the node with smallest  $f(n)=g(n)+h(n)$  (where  $g(n)$  = cost so far,  $h(n)$  = heuristic estimate to goal)
4. Uses: Priority queue (min-heap), heuristic function
5. Guarantees:
  - Optimal if the heuristic is admissible (never overestimates)
  - More efficient than Dijkstra when the heuristic is informative

# WHY A\* IMPROVES ON DIJKSTRA ?

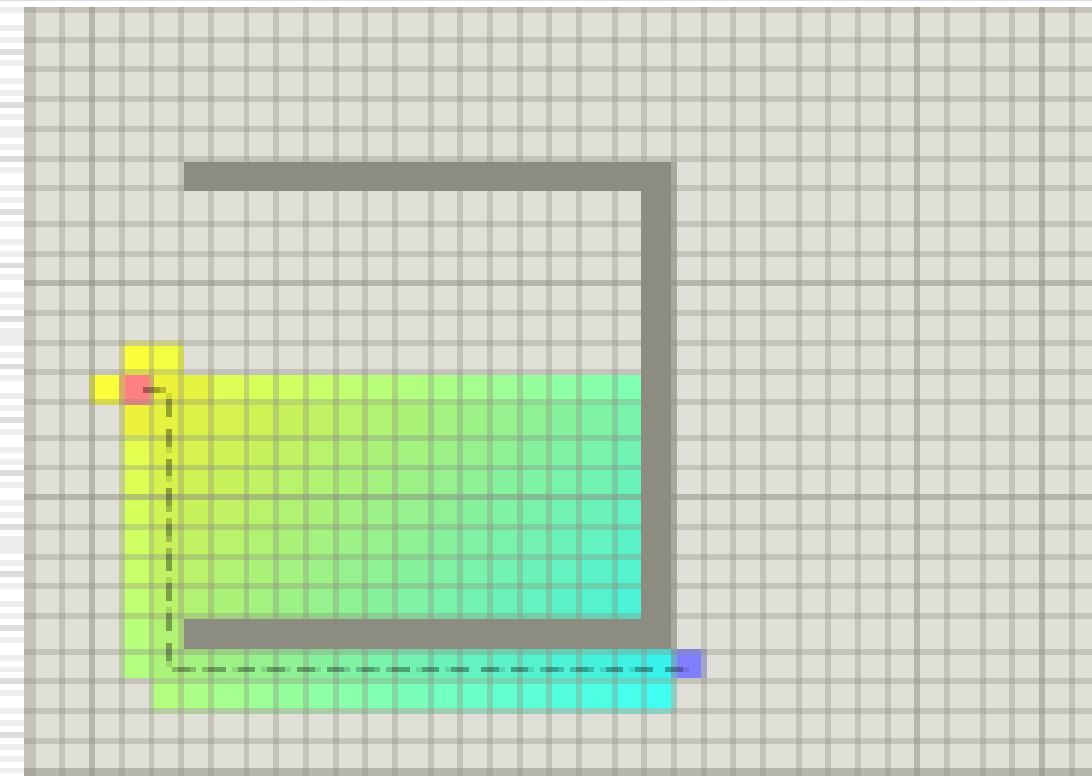
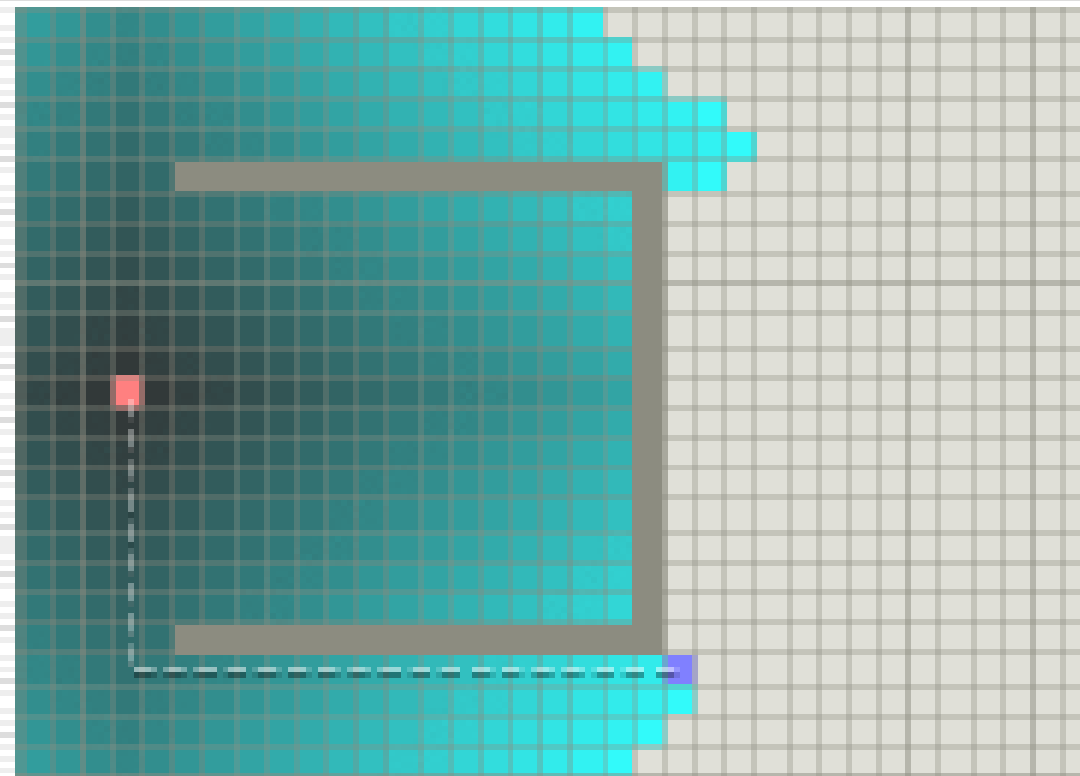
- Dijkstra explores uniformly in all directions
- Often unnecessary when goal is known
- A\* adds a heuristic:
  - Estimates distance to goal
  - Guides search efficiently

$$f(n) = g(n) + h(n)$$

Without  
obstacle



With  
obstacle



Dijkstra

A\*

# KEY RESULTS & TAKEAWAYS

- BFS & DFS: foundational but limited
- Dijkstra: optimal for weighted graphs
- A\*: optimal and efficient with good heuristics
- Grid planning is practical but approximate

Final message: Graph theory concepts directly power real-world path planning.



The background features abstract geometric elements. On the left, a large blue triangle points towards the center, with several thin, parallel white lines extending from its top-left corner. On the right, a large blue triangle points towards the center, with several thin, parallel white lines extending from its bottom-right corner. In the top-right corner, there is an orange triangle pointing towards the center, overlaid with a grid of small blue dots. In the bottom-left corner, there is an orange triangle pointing towards the center, overlaid with a grid of small blue dots. The main text is centered in a dark blue, serif font.

# THANK YOU

For your attention