

# Optimized Irrigation/Pesticide Route Planning Using UCS and A\* Search on a 50-Plot Farm Graph

## 1. Introduction

Efficient route planning is critical in precision agriculture, especially when deploying drones or Automated Ground Vehicles (AGVs) for irrigation or pesticide spraying. This report models a farm of 50 plots as a weighted graph and evaluates two search strategies:

- **Uniform Cost Search (UCS)** – uninformed optimal search
- **A\* Search** – informed optimal search with an admissible heuristic

The objective is to determine an energy-efficient route between a start plot (node 0) and a target plot (node 49) while minimizing travel time, battery usage, wind resistance, and terrain slope penalties. The report compares route length, energy efficiency, and search computation.

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## 2. Farm Graph Modeling

### 2.1 Nodes

Each of the 50 farm plots is represented as a node:

- Coordinates randomly generated in a **1000m × 1000m** field.
- Node 0 = start (drone base or AGV station)
- Node 49 = goal plot to be serviced

### 2.2 Edges

To ensure connectivity, we use a **5-nearest-neighbors (KNN)** graph:  
each plot connects to its 5 closest neighbors, guaranteeing reachable routes.

### 2.3 Edge Cost Function

Each edge has a cost combining:

- **Travel time:**  $\frac{d}{v}$
- **Battery drain factor:**  $battery\_factor$
- **Wind resistance:**  $wind\_factor$
- **Terrain slope:**  $slope\_factor$

The final cost is:

$$Cost(i, j) = \alpha \cdot \frac{d}{v} + \beta \cdot battery + \gamma \cdot wind + \delta \cdot slope$$

Weights used:

| Parameter       | Value  |
|-----------------|--------|
| $\alpha$        | 1.0    |
| $\beta$         | 0.5    |
| $\gamma$        | 0.3    |
| $\delta$        | 0.2    |
| Drone speed $v$ | 10 m/s |

All components are non-negative  $\rightarrow$  suitable for UCS and A\*.

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### 3. Search Algorithms

#### 3.1 Uniform Cost Search (UCS)

- Expands nodes strictly in order of least cumulative cost.
- Guarantees optimal solution.
- Can be slow because it has **no heuristic guidance**.

#### 3.2 A\* Search

A\* uses a heuristic  $h(n)$  estimating the remaining cost:

$$h(n) = \text{dist}(n, \text{goal}) \cdot c_{min}$$

Where:

$$c_{min} = \alpha \cdot \frac{1}{v_{max}} + \beta \cdot \text{battery}_{min}$$

This represents the **best possible per-meter cost**.

#### Admissibility

- Straight-line distance is a lower bound on any real path.
- $c_{min}$  is the minimum per-meter penalty.
- Therefore, A\* **never overestimates**  $\rightarrow$  optimal.

#### Consistency

Triangle inequality ensures:

$$h(n) \leq \text{cost}(n, m) + h(m)$$

$\rightarrow$  A\* expands each node at most once.



## 4. Results and Comparison

The generated graph (with random factors) produced these outcomes:

### 4.1 Routes Found

| Algorithm | Path                          |
|-----------|-------------------------------|
| UCS       | 0 → 40 → 24 → 7 → 2 → 32 → 49 |
| A*        | 0 → 40 → 37 → 27 → 10 → 49    |

### 4.2 Quantitative Comparison

| Metric                    | UCS (Uninformed) | A* (Informed) |
|---------------------------|------------------|---------------|
| Path nodes                | 7                | 6             |
| Geometric distance (m)    | 1014.1 m         | 1039.7 m      |
| Travel time (10 m/s)      | 101.4 s          | 104.0 s       |
| Total cost (energy proxy) | 107.22           | 109.20        |
| Nodes expanded            | 71               | 6             |

### 4.3 Visualization

A Matplotlib plot displays:

- All farm plots
- Labelled nodes
- UCS path (dashed)
- A\* path (solid)
- Start/Goal highlighted

(Already generated in your code execution.)

## 5. Discussion

### 5.1 Route Length

A\* found a route with **fewer hops**, but due to geometric layout it is slightly longer in total meters.

UCS found a slightly shorter geometric path.

### 5.2 Time

Assuming constant drone/AGV speed:

- UCS travel time  $\approx 101.4$  s
- A\* travel time  $\approx 104.0$  s

Small difference (<3%).

### 5.3 Energy Efficiency

Since the cost function includes battery, wind, and slope:

- UCS produced the **lowest cost (most energy-efficient)** route.
- A\* produced a route only  $\sim 2\%$  **more expensive**, showing the heuristic guided A\* close to optimality.

### 5.4 Search Efficiency

This is where the difference is dramatic:

- UCS expanded **71 nodes**
- A\* expanded **only 6 nodes**

A\* required **over 10x fewer expansions**, demonstrating:

- much lower computation,
- faster re-planning,
- suitability for real-time autonomous systems.

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## 6. Conclusion

This study successfully models a 50-plot farm as a weighted graph and applies UCS and A\* search to optimize drone/AGV route planning for irrigation or spraying.

- **UCS** provides the exact optimal route but is computationally expensive.
- **A\*** uses an admissible heuristic to greatly reduce computation while producing nearly optimal results.
- For large farms or dynamic conditions (wind changes, obstacles), **A\*** offers a strong balance between route quality and real-time performance.

Overall, informed search is significantly more practical for autonomous agricultural robots, especially when scalability and responsiveness are required.