

- Build a polygonal map → construct a visibility graph (VG) → run A\* (or Dijkstra) on that graph.
- Just remember: raw VG paths tend to skim obstacle corners, so you must "grow" obstacles by at least the robot radius (configuration-space inflation) before building the graph.

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Here's how I'd structure it end-to-end, with a data format that will make your life easy.

## 1) Data you give the planner

Instead of screenshots, export coordinates from GeoGebra. Two simple options:

CSV (recommended)

```
bash
type,id,label,x,y
vertex,poly1,A,0.0,0.0
vertex,poly1,B,2.0,0.0
vertex,poly1,C,2.0,1.0
vertex,poly1,D,0.0,1.0
# ... one row per vertex, polygons must be ordered CCW and closed implicitly
start,,S,-1.0, 0.5
goal,,G, 6.0, 0.5
```

JSON

```
json
{
  "polygons": [
    {"id": "poly1", "label": "box", "vertices": [[0,0],[2,0],[2,1],[0,1]]}
  ],
  "start": [-1.0, 0.5],
  "goal": [ 6.0, 0.5],
  "robot_radius": 0.09
}
```

Having an explicit `polygon_id` (or label) per vertex is perfect: it lets us reconstruct polygons unambiguously exactly as you suggested. Include `robot_radius` (or wheelbase half-width if you prefer) so we can inflate obstacles.

`type, id, label, x, y`

Ex: `vertex, poly 1 , A , 50.0 , 50.0`

`vertex, poly 2 , F , 40.0 , 40.0`

`start, start_pt, Start, 5.0, 95.0`  
`goal, goal_pt, Goal, 95.0, 5.0`

## 2) Pipeline (why + how)

1. Inflate obstacles by the robot radius ("grow" them) so that a point-mass path is safe for your real robot.
2. Collect candidate nodes = all (inflated) polygon vertices + start + goal.
3. Visibility test between every pair of nodes: a straight segment is valid iff it lies in free space and doesn't intersect any inflated obstacle edge (touching at a common vertex is OK if you allow grazing, but safer to require a tiny clearance  $\epsilon$ ).
4. Build graph: weighted undirected edges with Euclidean distance.
5. A\* with heuristic = straight-line distance. (Complete and optimal on this graph; A\* expands fewest nodes with an admissible heuristic.)
6. Output path as a polyline. (Later, hand the polyline to your path follower / local nav.)

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This lines up with the course notes (plan globally with A\*/Dijkstra, VG requires obstacle growth, then follow the path and pair with local avoidance).

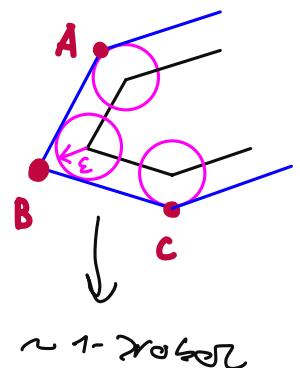
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## 3) Practical gotchas

- Numerical robustness: treat intersections with a small epsilon; snap nearly collinear points; ensure polygons are CCW and non-self-intersecting. ①
- Start/goal inside free space: verify they're not inside any inflated polygon. ②
- Performance: naive  $O(N^2)$  visibility checks are fine for coursework maps (tens–hundreds of vertices).
- Thymio reality check: VG gives shortest paths, but often too close to walls—keep the inflation generous to reduce controller stress.

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→ how much add  $\epsilon$



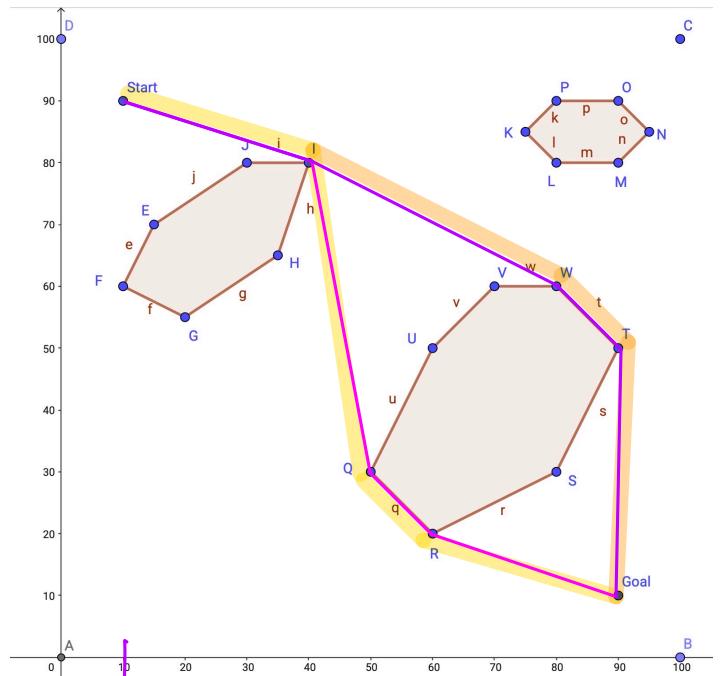
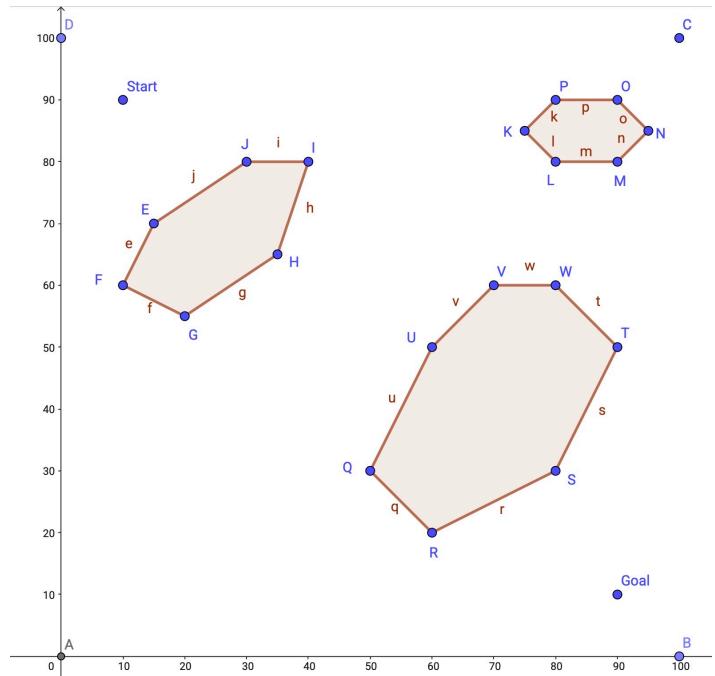
→ put hypothesis

→ would exist better than  $O(N^2)$ ?

Report: • hypothesis { CCW, non-self-intersecting }  
 { free space }

•  $O(N)$

• how determine  $\epsilon$



↓ pbm  $\alpha$

Python:

have coordinates  $\rightarrow$  draw polygons based edge coordinates?

list start nodes goal position angle

[ ]