

# Trijkstra

## A Dijkstra algorithm application to path planning

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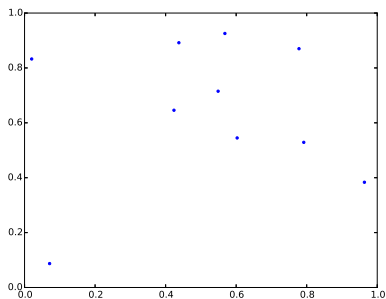


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# Voronoi diagrams

**Input:** A set of points in plane (or space) called **sites**

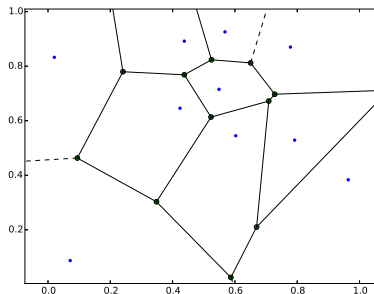
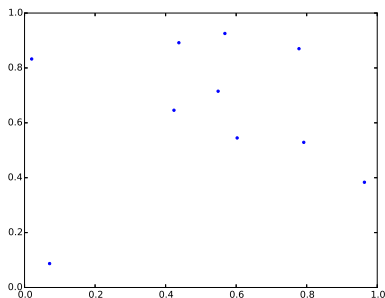
**Output:** A partition of the plane (or space) such that each point of a **region** is nearer to a certain site respect to the others



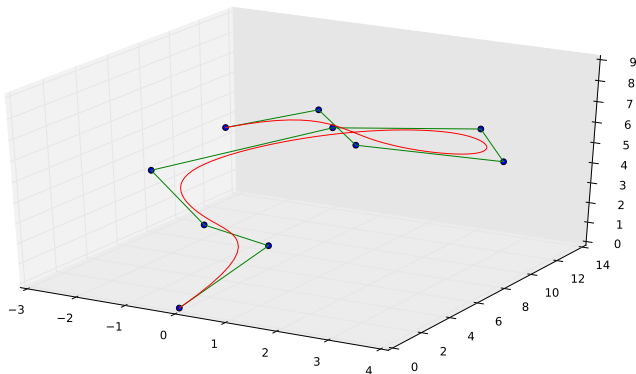
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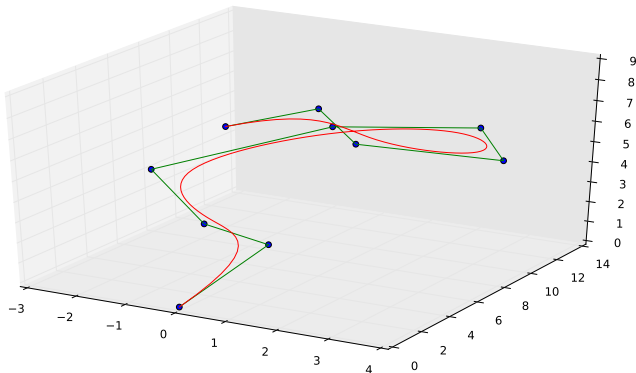


# B-spline



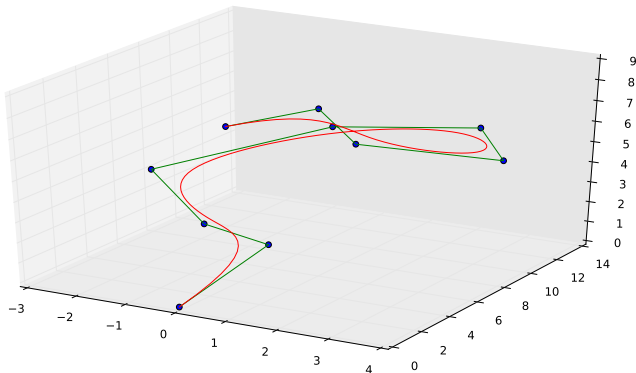
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- ✓ follow the shape of a control polygon
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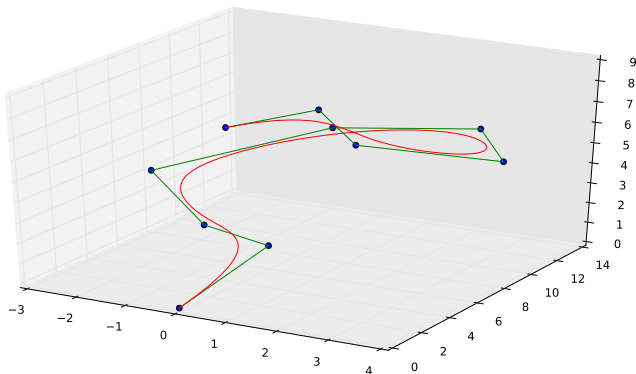
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# Dijkstra algorithm

```
1 def dijkstra(graph, start, end):
2     path = []
3     Q = priorityQueue.PQueue()
4     dist = {}
5     prev = {}
6     for node in graph.nodes(): #populate the queue
7         if node != start:
8             dist[node] = inf
9             Q.add(node, inf)
10        else:
11            dist[node] = 0
12            Q.add(node, 0)
13    while True: #main loop
14        u = Q.pop() #take nearest node and remove from queue
15        if u == end or dist[u] == inf: #finished (good or bad)
16            break
17        #all neighbors still in queue
18        for v in Q.filterGet(lambda node: node in graph.neighbors(u)):
19            tmpDist = dist[u] + graph[u][v]['weight']
20            if tmpDist < dist[v]: #if distance shorter update values
21                dist[v] = tmpDist
22                prev[v] = u
23                Q.add(v, tmpDist) #update distance also in queue
24    u = end
25    while u in prev: #backward recreation of path
26        u = prev[u]
27        path[:0] = [u]
28    if path:
29        path[len(path):] = [end]
30        path[:0] = [start]
31    return path
```



# Background

## Main problem

Path planning from a **start** point to an **end** point in 3D space with obstacles using **Voronoi** diagrams.

1. Distribute **points** in the surfaces of obstacles
  - ▶ and optionally in the surface of bounding box
2. Build **Voronoi** diagram using those points as source
3. Transform the Voronoi diagram in a **graph**
  - ▶ cells **vertexes** as **nodes**
  - ▶ cells **edges** as **arcs** (infinite edges ignored)
4. **Prune** the arcs that crosses an obstacle's surface
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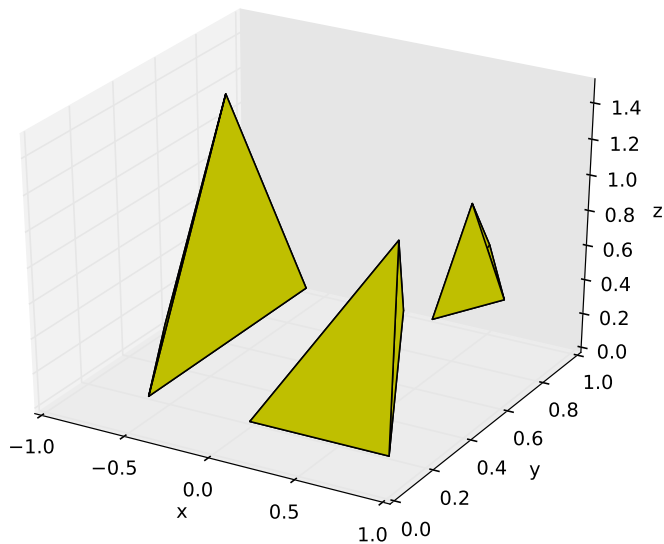
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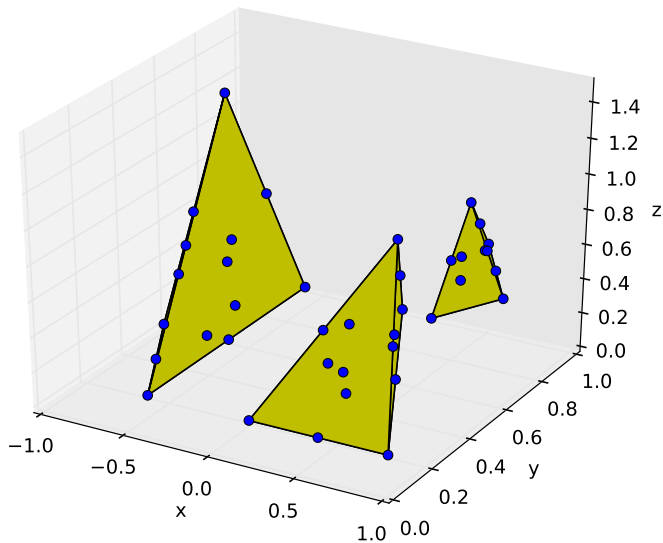
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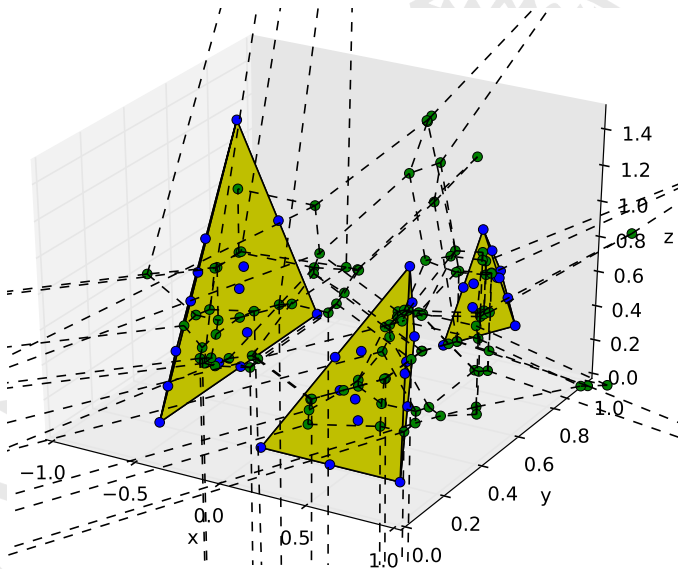




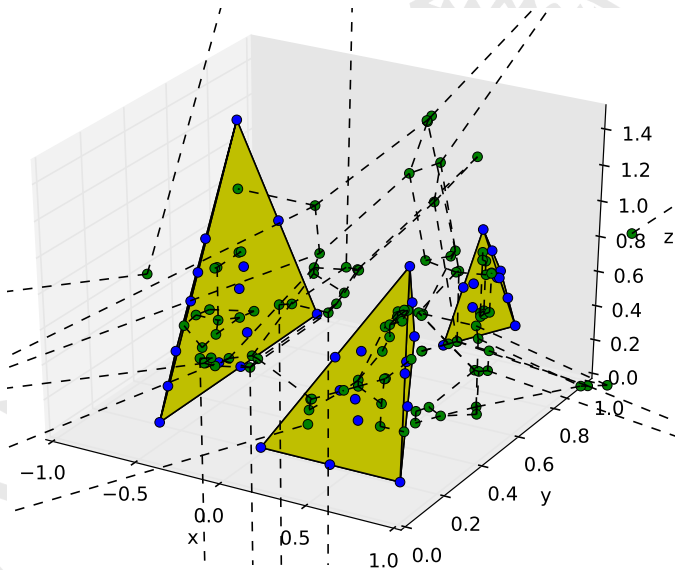
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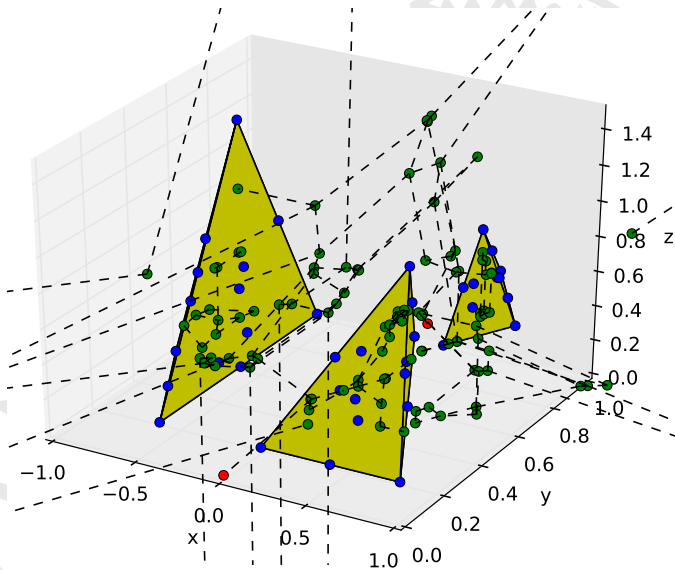
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## Idea

Make a **smoother** curve instead of finding the polygonal chain of the shortest path in the structure

- ✓ we can use a **B-Spline** that
  - ▶ **interpolate** the start and end vertexes
  - ▶ use the shortest path found with Dijkstra as **control polygon**

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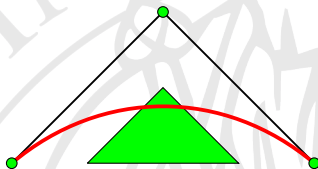
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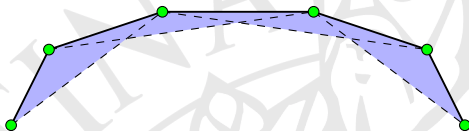


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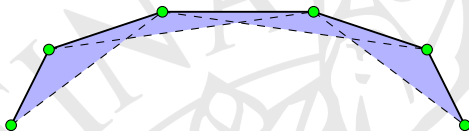


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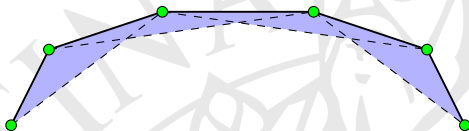


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7. the shortest **path** (with free triangular convex hull) can be obtained following the **previous** pointer from the **ending** triple, and **deconstructing** the triples

# Declarations & Triples creation

```
1 def _trijkstra(self, startA, endA):
2     start = tuple(startA)
3     end = tuple(endA)
4     endTriplet = (end, end, end) #special triplet for termination
5     inf = float("inf")
6     path = []
7     Q = priorityQueue.PQueue()
8     dist = {}
9     prev = {}
10    hits = []
```

```
1 for node0 in self._graph.nodes():
2     for node1 in self._graph.neighbors(node0):
3         for node2 in filter(lambda node: node!=node0, self._graph.neighbors(node1)):
4             triplet = (node0, node1, node2)
5             if not triplet[:-1] in hits:
6                 if not self._triangleIntersectPolyhedrons(np.array(node0), np.array(node1),
7                     ↪ np.array(node2)):
8                     if node0 != start:
9                         dist[triplet] = inf
10                        Q.add(triplet, inf)
11                    else:
12                        dist[triplet] = 0
13                        Q.add(triplet, 0)
14                else:
15                    hits[:0] = [triplet]
16
17    dist[endTriplet] = inf
18    Q.add(endTriplet, inf)
```

# Main loop

```
1 while True:
2     u = Q.pop()
3
4     if u == endTriplet or dist[u] == inf:
5         break
6
7     for v in Q.filterGet(lambda tri: u[1] == tri[0] and u[2] == tri[1]):
8         tmpDist = dist[u] + self._graph[u[0]][u[1]]['weight']
9         if tmpDist < dist[v]:
10             dist[v] = tmpDist
11             prev[v] = u
12             Q.add(v, tmpDist)
13
14     if u[2] == end:
15         tmpDist = dist[u] + self._graph[u[0]][u[1]]['weight'] +
16             ↪ self._graph[u[1]][u[2]]['weight']
17         if tmpDist < dist[endTriplet]:
18             dist[endTriplet] = tmpDist
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# Path creation

```
1 u = endTriplet
2 while u in prev:
3     u = prev[u]
4     path[:0] = [u[1]]
5
6 if path:
7     path[len(path):] = [end]
8     path[:0] = [start]
9
10 return np.array(path)
```

## After

we can use the returned `path` as a `control polygon` for a quadratic B-Spline without problems, and construct a smoother path.

# Path creation

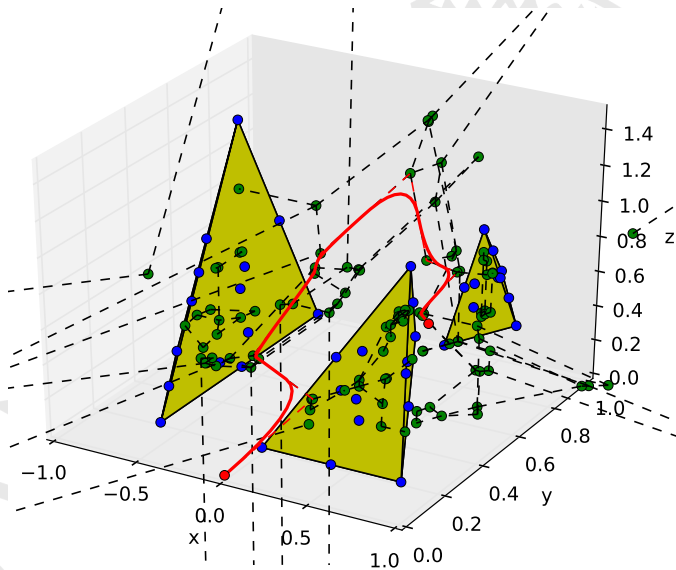
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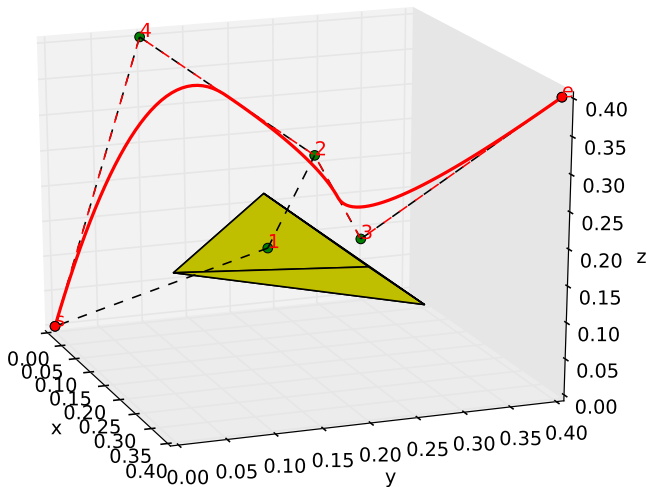
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## Previous example



# Clearer example



# Complexity considerations

The algorithm is analogous to classical **Dijkstra** applied to a **transformed** graph where:

- ✓ the original graph is **not** directed and weighted
- ✓ the transformed graph is **directed** and weighted, and
  - ▶ if  $A$  and  $B$  are **neighbouring** and  $B$  and  $C$  are **neighbouring**, in the original graph
  - ▶ we have two nodes  $(A, B, C)$  and  $(C, B, A)$  in the transformed graph
  - ▶ a node  $(A_1, B_1, C_1)$  is a predecessor of  $(A_2, B_2, C_2)$  in the transformed graph if  $B_1 = A_2$  and  $C_1 = B_2$  in the original graph
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# Time complexity (For original graph creation)

- ✓ **Fortune** algorithm run in  $\mathcal{O}(n \log n)$
- ✓ if we have  $n$  input sites we get  $\mathcal{O}(n)$  vertexes of Voronoi areas
- ✓ if we assume obstacles with a maximum area
  - ▶  $\mathcal{O}(|Ob|)$  input sites

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✓  $\mathcal{O}(k^3|V_{orig}| + k^2|V_{orig}|\log(k|V_{orig}|))$

If  $k$  is constant (don't grow with  $|V_{orig}|$ )

- ✓ Total cost:  $\mathcal{O}(|V_{orig}|\log|V_{orig}|)$
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If the graph is a clique

- ✓  $k = |V_{orig}| - 1$
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Total (if obstacle area is not a function of the number of obstacles)

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- ✓ The predominant factor is for checking the triples collisions with obstacles
  - ▶ if  $k$  is constant  $\mathcal{O}(|Ob|^2)$
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