Dijkstra’s algorithm,
Fast marching & Level set &

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Looking back

• Medical image segmentation is (usually) selecting a suitable method from a toolbox of available approaches

• Making problem dependent customizations improvements to include a priori information
Tools in the toolbox

- Statistical methods (Markov fields etc)
- Active contours
- Dynamical programming
- Fast marching
- Level set methods
- Variational methods
- Chan-Vese segmentation
- .....
Purpose of this lecture(s)

- Dijkstra’s shortest path algorithm
- Understanding of fast marching
- Understanding of level set method
- Introducing task 3
Dijkstra’s - motivation

Radial Resample and filtering

Shortest path algorithm on edge image
Dijkstra’s - motivation

Boundary tracing of ice radar profile on Svalbard glacier
Mathematical formulation

\[
\min \sum_{<(i,j),(k,l)> \in A} c_{(i,j)(k,l)} x_{(i,j)(k,l)}
\]

\[
\begin{aligned}
&x_{(i,j)(k,l)} \in \{0,1\} \\
&\sum_{(i,j) \in S(i)} x_{(i,j)(k,l)} - \sum_{(i,j) \in T(i)} x_{(k,l)(i,j)} = \\
&\begin{cases}
-1 & \text{if } (i,j) = S \\
1 & \text{if } (i,j) = T \\
0 & \text{otherwise}
\end{cases}
\end{aligned}
\]

Where \(<(i,j),(k,l)>\) is a vertex, \(c_{(i,j)(k,l)}>0\) is the cost for the vertex, \(x_{(i,j)(k,l)}\) is 1 if the vertex is included in the optimal path, \(S\) is the starting point, and \(T\) is the end point.
Dijkstra’s algorithm

1. Start with node S
2. Keep track of a set of selected nodes $N$ (black)
3. Select the shortest vertex $w$ out of $N$
4. Keep track of precessor
5. Repeat until all nodes in $N$
Dijkstra’s algorithm

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5. Repeat until all nodes in \( N \)
Dijkstra’s algorithm

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5. Repeat until all nodes in $N$
Practical implementation

1. Keep a list of selected nodes ($N$)
2. Store a sorted list of outward vertices
3. Take lowest cost vertex from list, add node to $N$
4. Add new vertices to the sorted list
5. Repeat from 3) until all nodes are selected
6. Shortest path is “backtracked”

Suitable data structure for the sorted list is a sorted heap.
Optimality

- We keep track on selected nodes $N$

- In each iteration find the shortest vertex $w$ that connects to a node outside of $N$

- Need to prove that there can not be a shorter path from start to the node connected by the shortest vertex $w$
Optimality

Suppose a such path exists

1. \(w\) was selected as the shortest vertex in \(N\) to the node
2. Cost between nodes are \(\geq 0\).
3. Hypothetical shorter path must go via an intermediate node as \(w\) was chosen as the smallest vertex from the set \(N\).
4. The hypothetical path can not be shorter than the path to \(w\) as costs \(\geq 0\). Thus the calculated path is the optimal shortest path.
Dijkstra’s algorithm

Artificial startpoint S

Artificial endpoint T

Nodes = pixels

Rows in image

Columns in image
Dijkstra’s algorithm

Artificial startpoint S

Artificial endpoint T
Dijkstra’s algorithm

Artificial startpoint S

Artificial endpoint T
Dijkstra’s algorithm

Artificial startpoint $S$

Artificial endpoint $T$
Dijkstra’s algorithm
Dijkstra’s algorithm
Dijkstra’s algorithm

Artificial startpoint S

Artificial endpoint T
Continue....
Dijkstra’s algorithm
Dijkstra’s algorithm

Artificial startpoint S

Artificial endpoint T

Final optimal path
Setting the costs

- Problem dependent
- Low cost for favorable areas?
- Low cost for edges?
- Penalize large jumps in image (derivative)
- Penalize curvature (how....?)
- Force a closed line?
**Interface propagation**

Outside properties, such as curvature & normal direction

Inside properties

Outside properties, for instance underlying vector field

Speed $F(L, G, I)$

$L = \text{Local properties, such as curvature & normal direction}$

$G = \text{Global properties}$

$I = \text{Independent properties, for instance underlying vector field}$
Representing moving interfaces as “height curves”

Advantages:
- Variable topology
- Generic

Disadvantages:
- Difficult to introduce a priori to constrain topology
Object representation
Advantages

- Expansion to higher dimensions trivial
- Topological changes naturally
- Simple to calculate curvature
- Efficient computational schemes can be developed
Different interface formulations

- Boundary value formulation
- Initial value formulation
Boundary value formulation

- Assume $F>0$ (i.e., always outward motion)
- Describe interface by computing arrival time $T(x,y)$
- Use distance = rate*time

\[
dx = FdT \\
1 = F \frac{dT}{dx}
\]
Boundary value formulation

• In higher dimensions $dT$ becomes $\nabla T$ which gives the Eikonal equation:

$$|\nabla T|F = 1$$
Examples
Examples

Speed image

Startpoint

Arrival time
Examples

- Speed image
- Arrival time

Startpoint
Examples

Speed image

Arrival time
Key issues

- Finding good starting point(s) where $T=0$
- Finding suitable speed map (translating intensity and edges)
- Selecting the object based on arrival time
Initial value formulation

Assume that $F$ can be both positive and negative.

**Problem:** The boundary can pass a certain position multiple times.

**Solution:** Represent the curve as the zero level of a time dependent level set function $\varphi$ instead of arrival time
Initial value formulation

The level set value for a point on the boundary must always be zero:

$$\phi(x(t), t) = 0$$

Chain rule gives:

$$\phi_t + \nabla \phi(x(t), t) \cdot \frac{dx(t)}{dt} = 0$$

We have \( \frac{dx(t)}{dt} \cdot \hat{n} = F \), \( \hat{n} = \nabla \phi / |\nabla \phi| \) gives:

$$\phi_t + F|\nabla \phi| = 0$$
Boundary value formulation

\[ |\nabla T|F = 1 \]

Front representation:
\[ \Gamma(t) = \{(x, y) T(x, y) = t\} \]

Requires \( F > 0 \)

Advantage: Fast

Initial value formulation

\[ \phi_t + F|\nabla \phi| = 0 \]

Front representation:
\[ \Gamma(t) = \{(x, y) \phi(x, y, t) = 0\} \]

Arbitrary \( F \)

Advantage: Can use shape dependent speed function (curvature, smoothing)
Fast numerical solution to Eikonal equation - Fastmarching

Solution similar to Dijkstra’s:

1. For each neighbor to the starting node(s) compute arrival time. Store arrival times in a sorted list of arrival times.

2. Choose the next unselected node with the shortest arrival time.

3. For each unselected neighbor compute arrival time. Store arrival times in the sorted list of arrival times.

4. Repeat from 2 until all nodes selected (or another stop criteria).
Calculating arrival time

- Uses arrival time of selected nodes (upwind scheme)
- Use hyperbolic conservation laws =>

\[
|\nabla T|^2 = \begin{cases} 
\max(V_A - V_B, V_A - V_C, 0)^2 + \\
\max(V_A - V_D, V_A - V_E, 0)^2 + \\
\max(V_A - V_F, V_A - V_G, 0)^2 
\end{cases}
\]

where \( V \) is distance for nodes, and \( V_A \) is sought.

- Solve by 1) ignoring unselected nodes, 2) reduce by looking at \( V_B < V_C \? , V_D < V_E \? , V_F < V_G \? \)

- Remains a second order equation to solve where largest value is taken (if two solutions exist). Ex \((V_A - V_B)^2 + (V_A - V_E)^2 = F^2\)

- Information flows from selected nodes (“upwind”) to unselected nodes (“downwind”, and towards higher arrival times).
Fast numerical solution to initial value formulation – Narrow band

Solve equation in a narrow band (grey pixels) close to the zero level set. Fastmarching can be used for initialization.
Time complexity

• Dijkstra’s $O(E+N \log N)$ where $E$ is the number of edges in the graph and $N$ is the number of nodes.

• Fast marching $O(N \log N)$ where $N$ is the number of nodes.

• This assumes heap implementation for managing the sorted list.
Take home messages

• Dijkstra’s solves shortest path in an optimal sense.

• Boundary formulation / initial value formulation.

• Fast numeric schemes using sorted heap.

• Trick is selecting good speed maps and finding stop criteria.
Short about hand-in assignment

- DICOM reader
- Display medical images
- Fastmarching in 2D and 3D