

# Exact and Approximate Shortest Paths

Siddhartha Chaudhuri

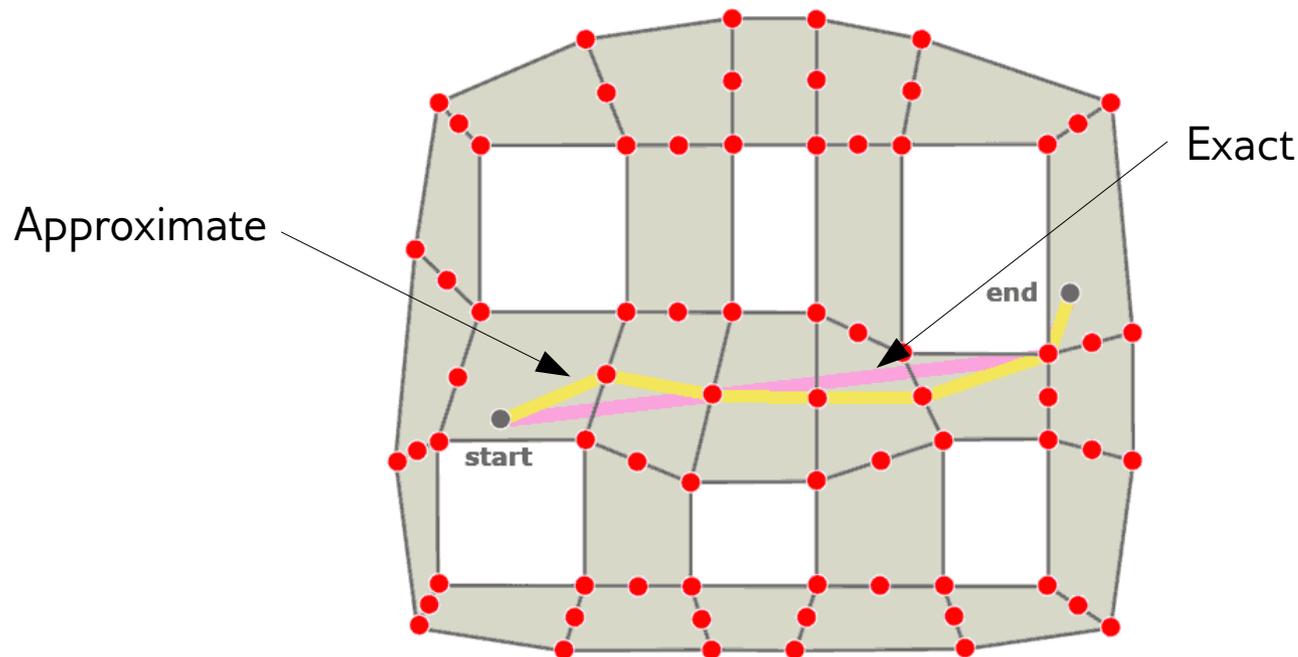
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# Overview

- Last class:
  - Distances on surfaces: applications and complications
  - Definition of geodesics
    - ≠ shortest paths, but often conflated!
- Today:
  - An algorithm for finding exact shortest paths on a mesh
  - Can be modified to quickly give approximately shortest paths

# Challenge

- The shortest path traverses **interiors** of triangles
  - Where does the path cross edges?
  - Continuous optimization problem, much harder than shortest paths on discrete graphs



# A historical perspective

- Exact:

- Single source, all destinations

- Mitchell/Mount/Papadimitriou [1987]:  $O(n^2 \log n)$
- Chen/Han [1996]:  $O(n^2)$

- Single source, single destination

- Kapoor [1999]:  $O(n \log^2 n)$

- Approximate:

- Insert extra edges: Lanthier [1997]

- Iterative optimization: Kanai/Suzuki [2001], Martinez et al. [2004]

- Fast-marching: Kimmel/Sethian [1998,  $O(n \log n)$ ]

- Fast-sweeping: Zhao [2005,  $O(n)$ ]

- Window merging: Surazhsky et al. [2005,  $O(n \log n)$ ]

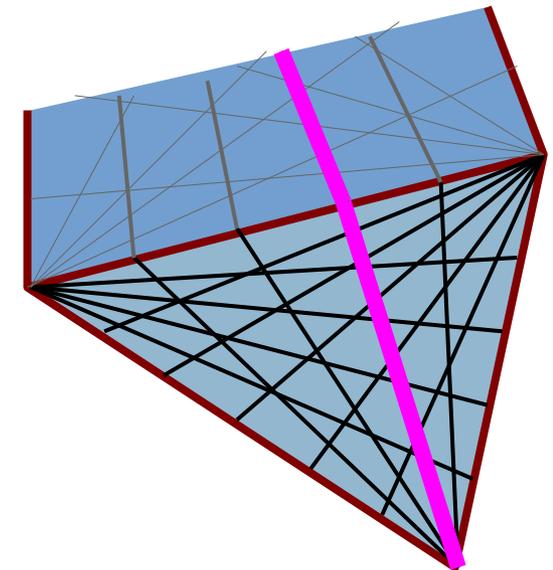
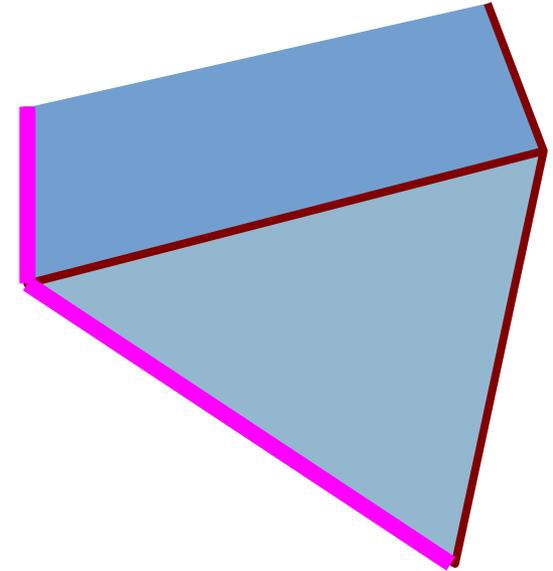
- Heat flow: Crane et al. [2013]

$n$  = number  
of triangles  
in the mesh

Note that asymptotic  
behavior  $\neq$  real-world  
performance

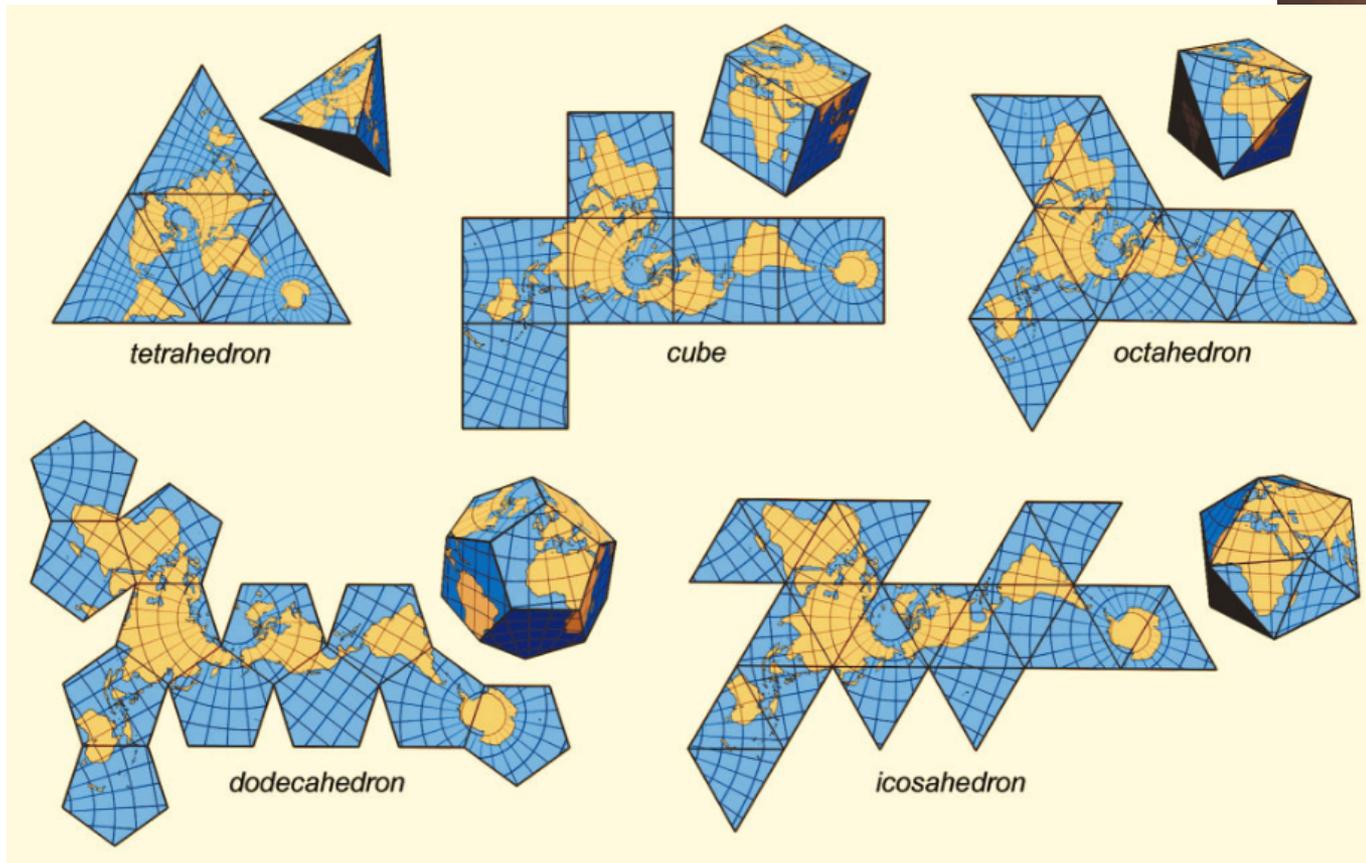
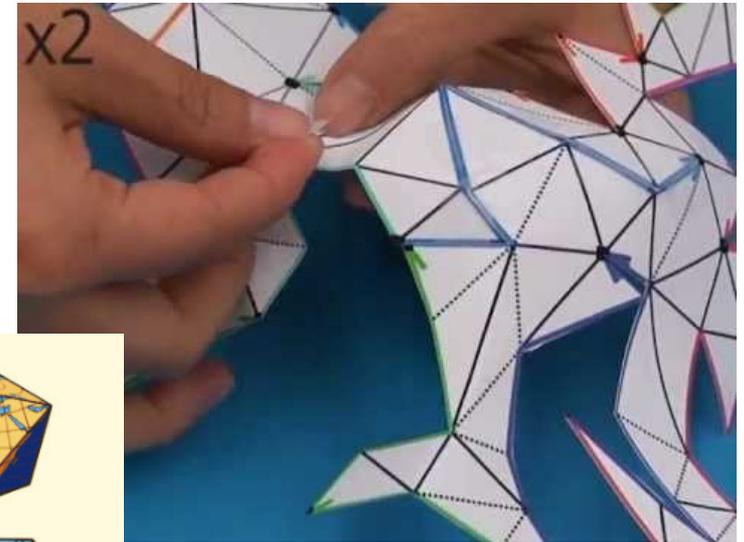
# Edge Insertion

- **Problem:** Edges crudely model surface connectivity
- **Solution:** Add more edges!
  - Finer-grained discrete problem
  - Produces better approximations with more edges, at the cost of runtime complexity
    - But typically converges at around 6 points per edge



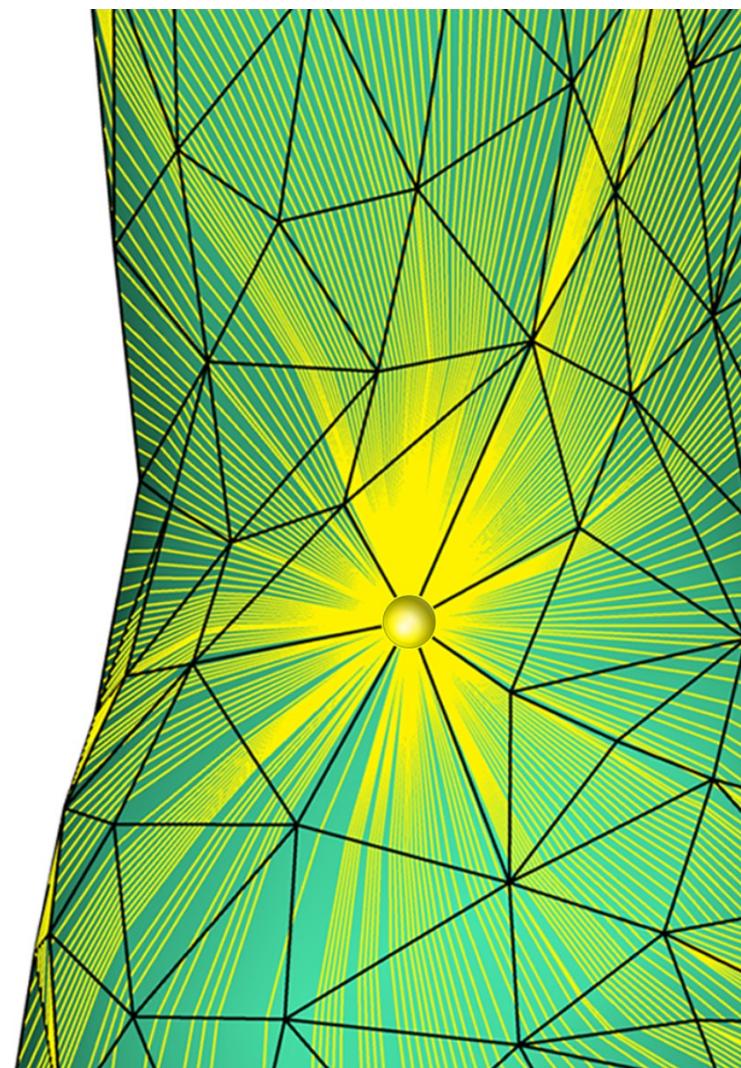
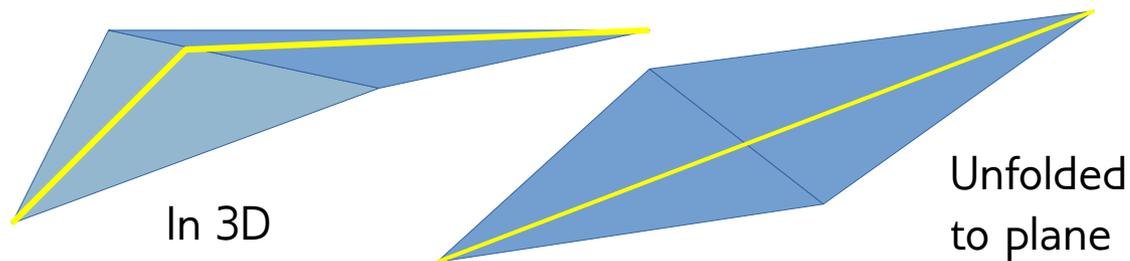
# A crucial observation 1/2

- A mesh can be **unfolded**

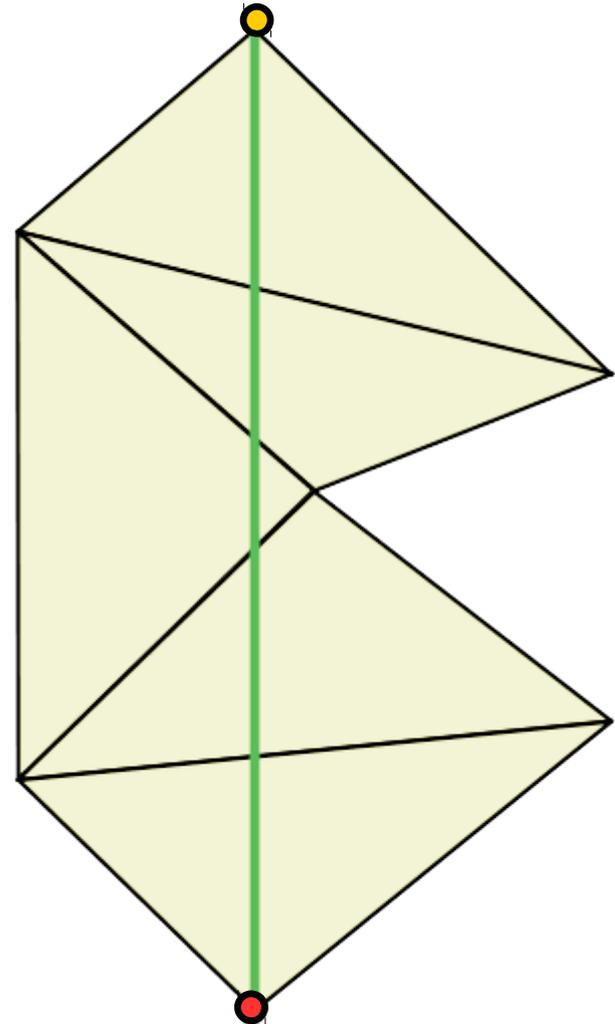
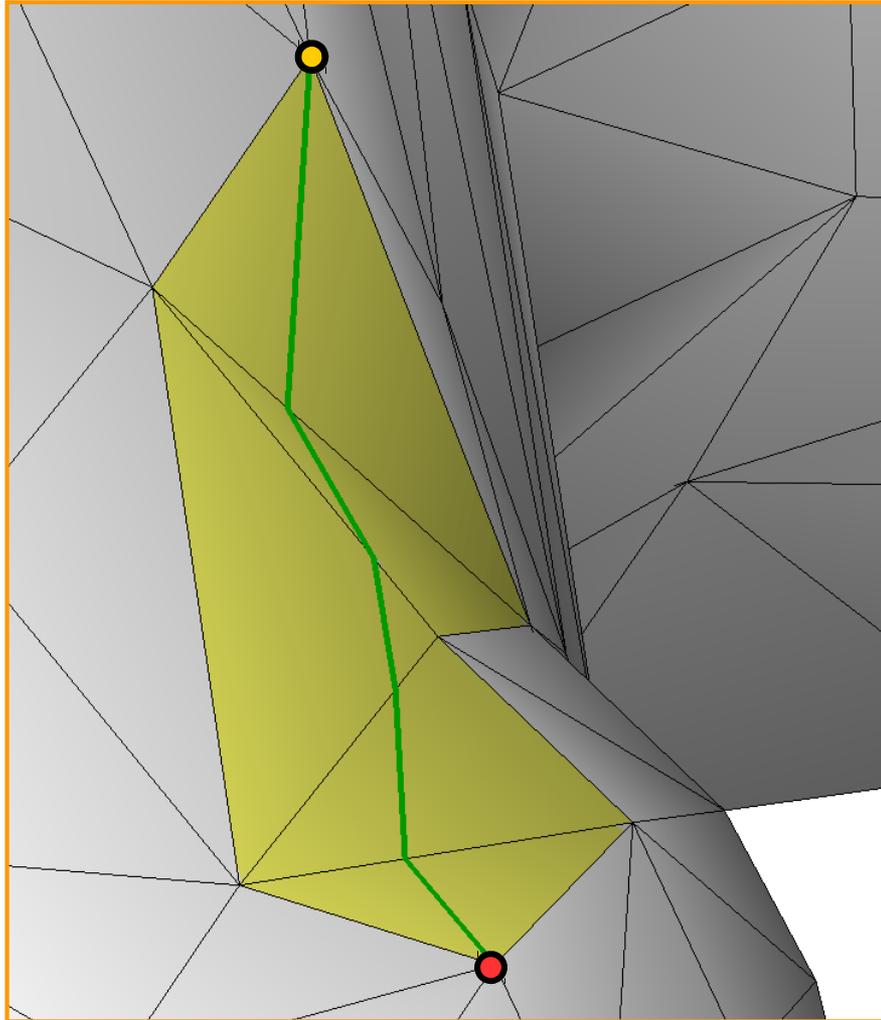


# A crucial observation 2/2

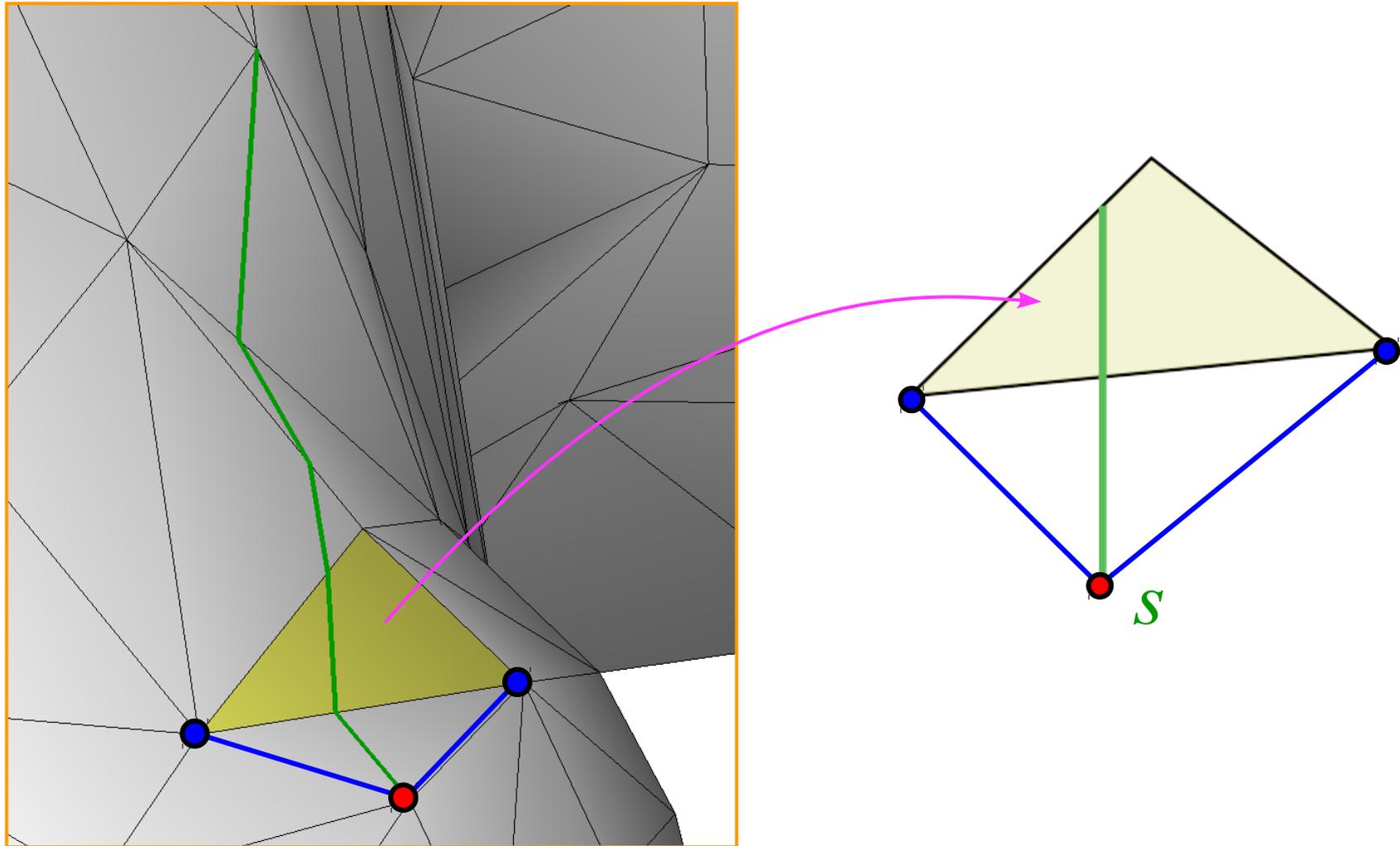
- Shortest paths can be visualized as rays emanating from point in all directions
  - Interior to triangle: shortest path must be straight line
  - Crossing edge: shortest path corresponds to straight line when two triangles are **unfolded** into common plane.



# Unfolding the shortest path

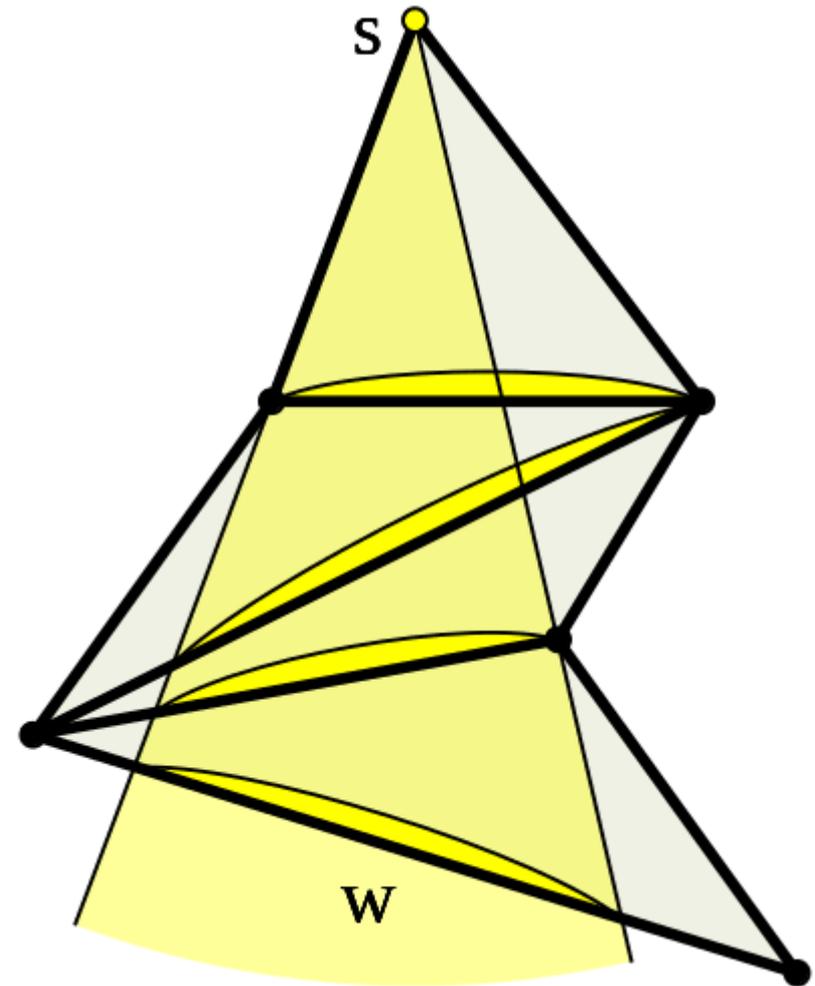


# Source vertex unfolded to triangle plane



# Windows

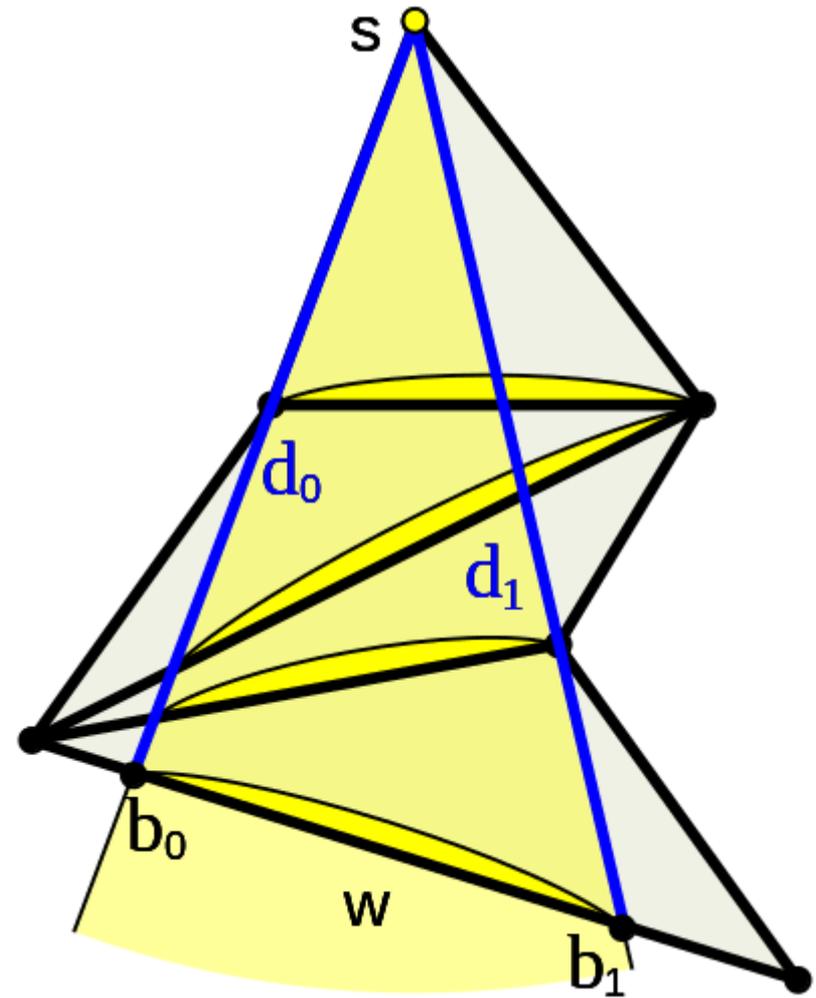
- A **window** is a segment of an edge over which all shortest paths to the source traverse the **same sequence of faces**
- Within a window, distance computations can be performed **atomically** (no need to worry about routing in the mesh)



Unfolded layout

# Window specified by 5-tuple

- $b_0, b_1$ : local  $x$ -coordinates of endpoints on edge
- $d_0, d_1$ : distances from endpoints to source  $S$
- $\tau$ : direction to source (side of edge where  $S$  lies)



Unfolded layout

# Source reconstruction

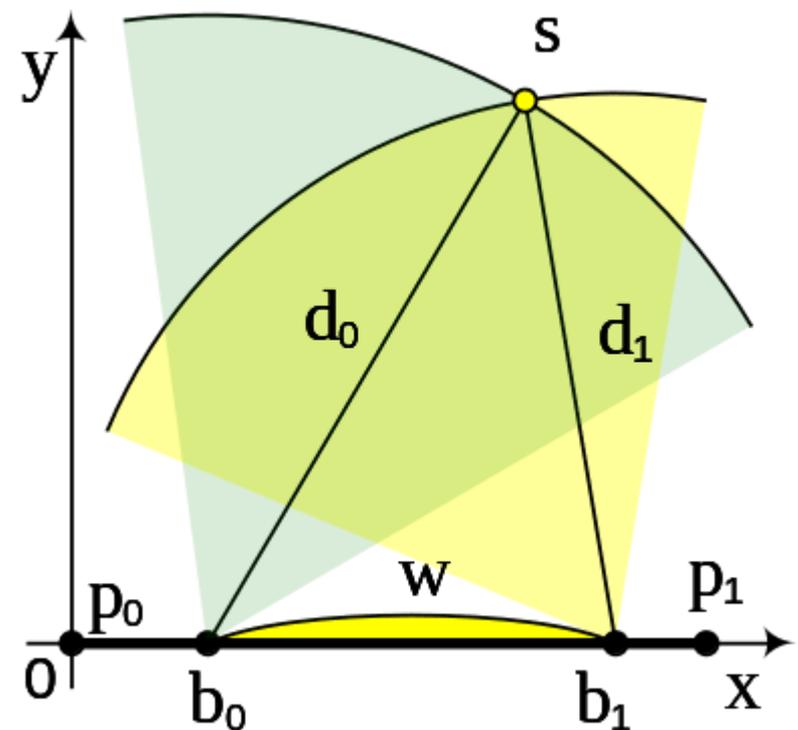
- Given two distances  $d_0$ ,  $d_1$ , recover the source  $s = (x, y)$
- Computation is simple via local coordinate system

$$(x - b_0)^2 + y^2 = d_0^2$$

$$(x - b_1)^2 + y^2 = d_1^2$$

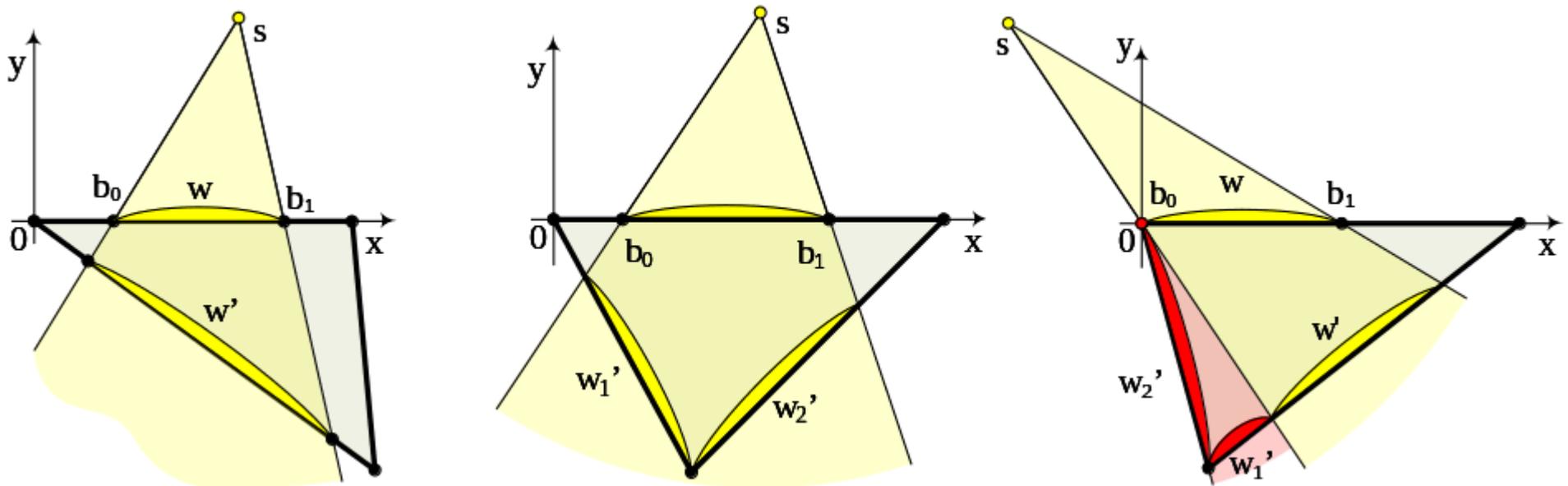
$$x = \frac{d_0^2 - d_1^2 - b_0^2 + b_1^2}{2(b_1 - b_0)}$$

$$y = +\sqrt{d_0^2 - (x - b_0)^2}$$



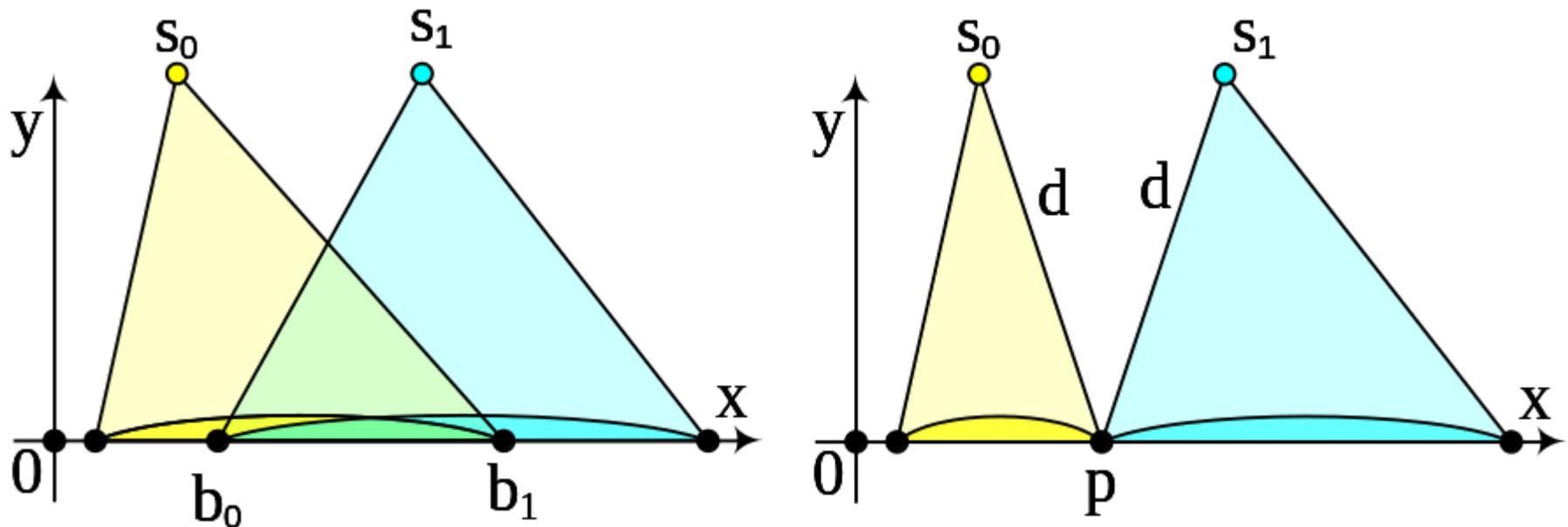
# Basic Idea: Window Propagation

- Step from triangle to adjacent triangle
- Windows on an edge create new windows on other edges of new triangle
  - The cone “sees” new edges as we enter the new triangle
- Can create one, two or three new windows



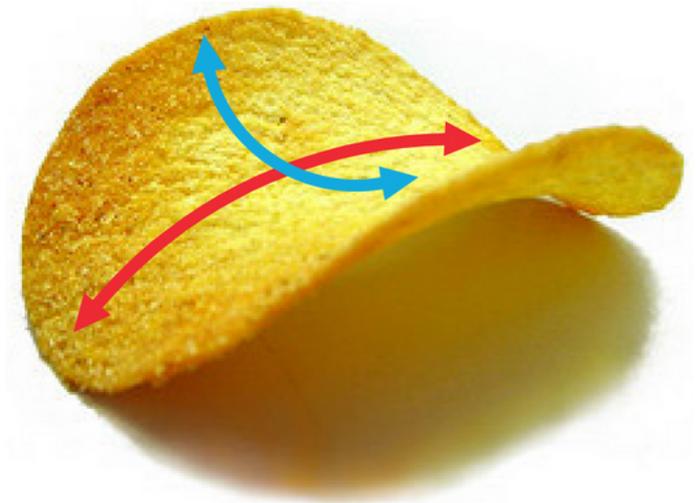
# Overlapping Windows

- Find equidistant point on edge
- Cut off overlapping parts that define larger distances
- Distance function is continuous on edge

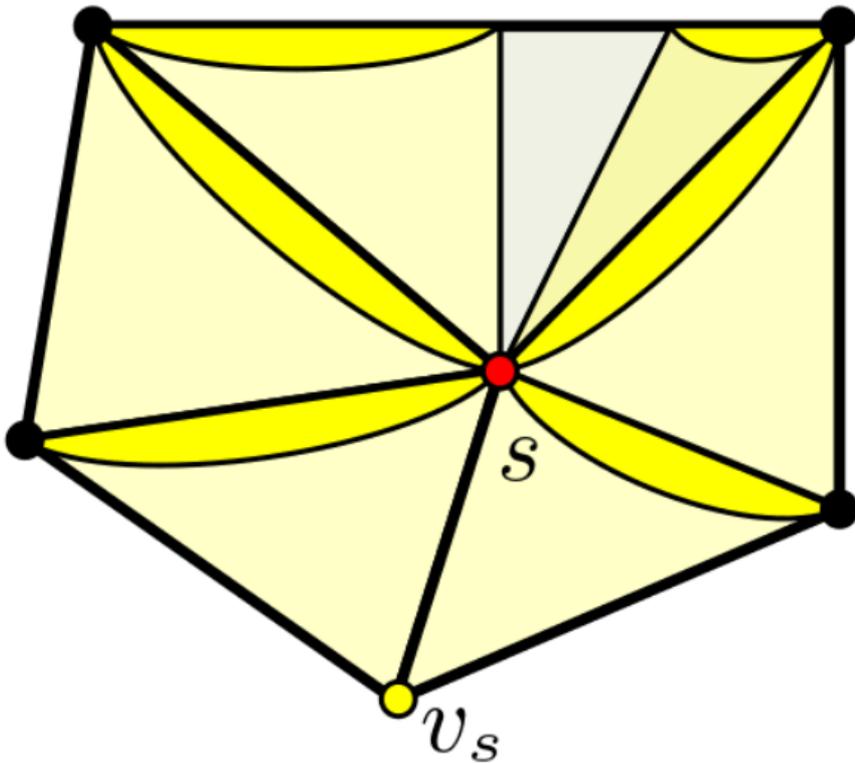


# Hyperbolic (saddle) vertices

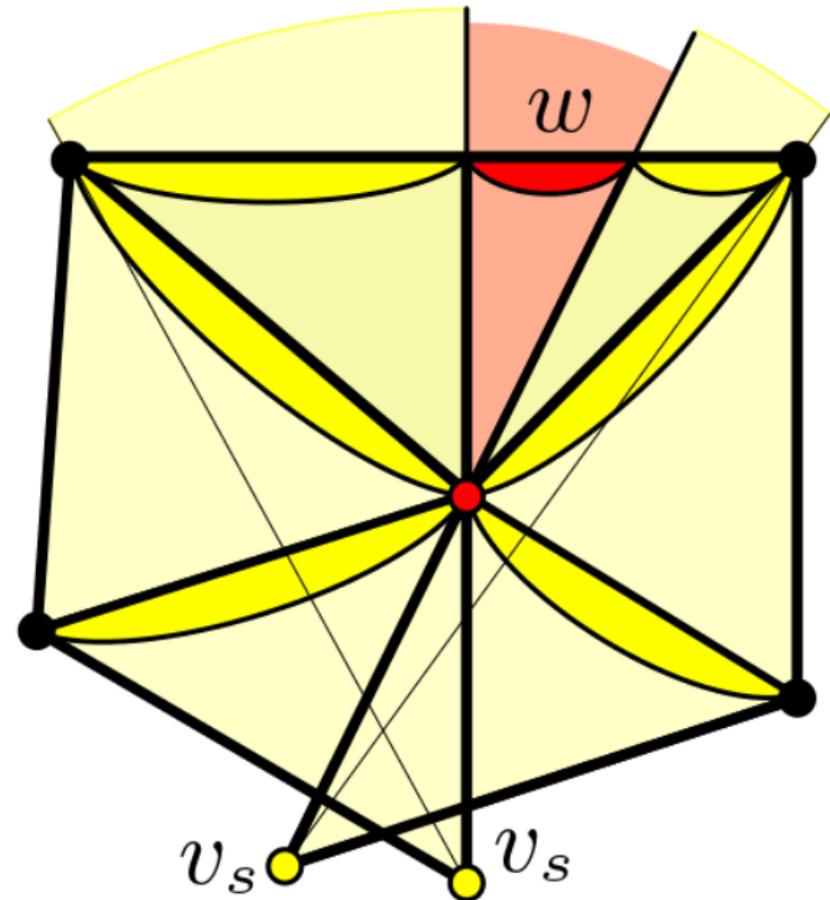
- If the sum of face angles at a vertex is  $> 2\pi$ , it is called a hyperbolic/saddle vertex
- It cannot be unfolded onto a plane without foldovers (overlapping faces)
- The shortest path can pass through **boundary**, **hyperbolic** ( $> 2\pi$ ) and **parabolic** ( $= 2\pi$ ) vertices
  - Hyperbolic vertices need special handling



# Hyperbolic (saddle) vertices



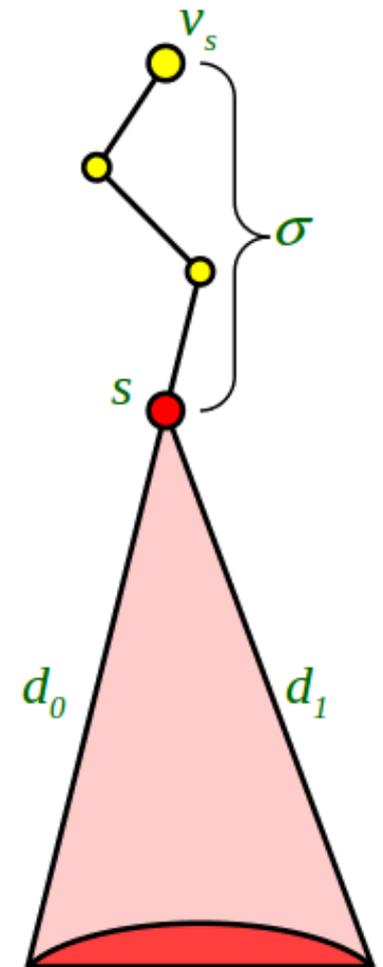
Unflattened region near saddle vertex  $s$ . Part of the upper triangle is not visible by rays from source vertex  $v_s$



Unfolding to plane of upper triangle reveals two different images of  $v_s$ , neither of which is visible from red region. All shortest paths to  $w$  pass through saddle vertex.

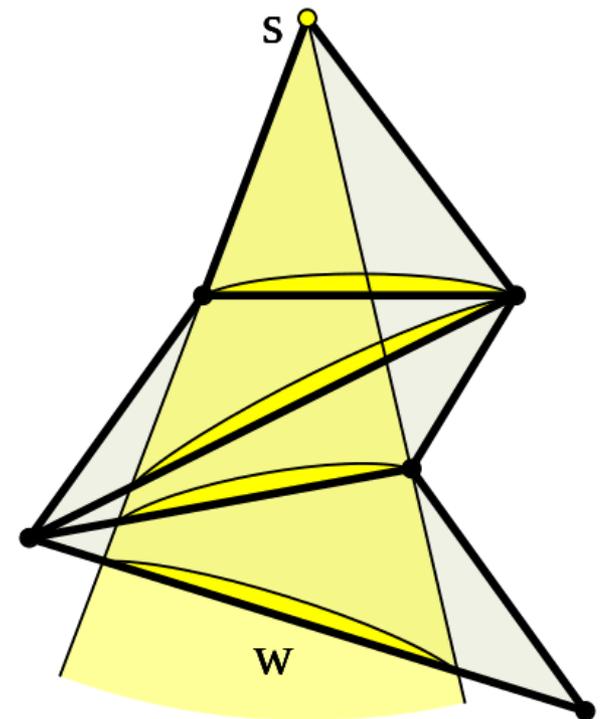
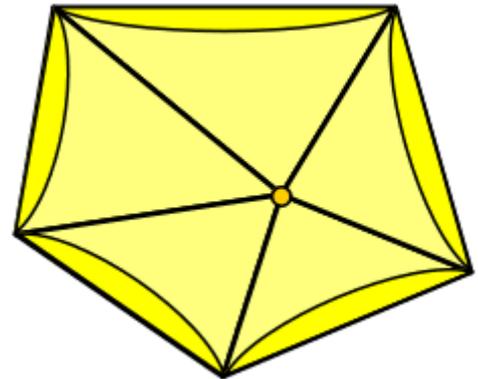
# Hyperbolic (saddle) vertices

- **Solution:** Treat saddle (or boundary) vertex as **pseudo-source**
  - All shortest paths route through pseudo-source
  - ... so we can originate our visibility cones from the pseudo-source, and add its distance from the actual source
- Window as **6-tuple**:  $\{ b_0, b_1, d_0, d_1, \sigma, \tau \}$ 
  - $\sigma$  : distance from **pseudo-source** to source



# The algorithm

- Initialize queue  $Q$  with a window for each edge adjacent to source  $s$ , sorted by distance to source
- Until  $Q$  is empty
  - **select** (and remove) a window from  $Q$
  - **propagate** selected window
  - **update**  $Q$  with new windows
- The algorithm fully covers each edge with non-overlapping windows

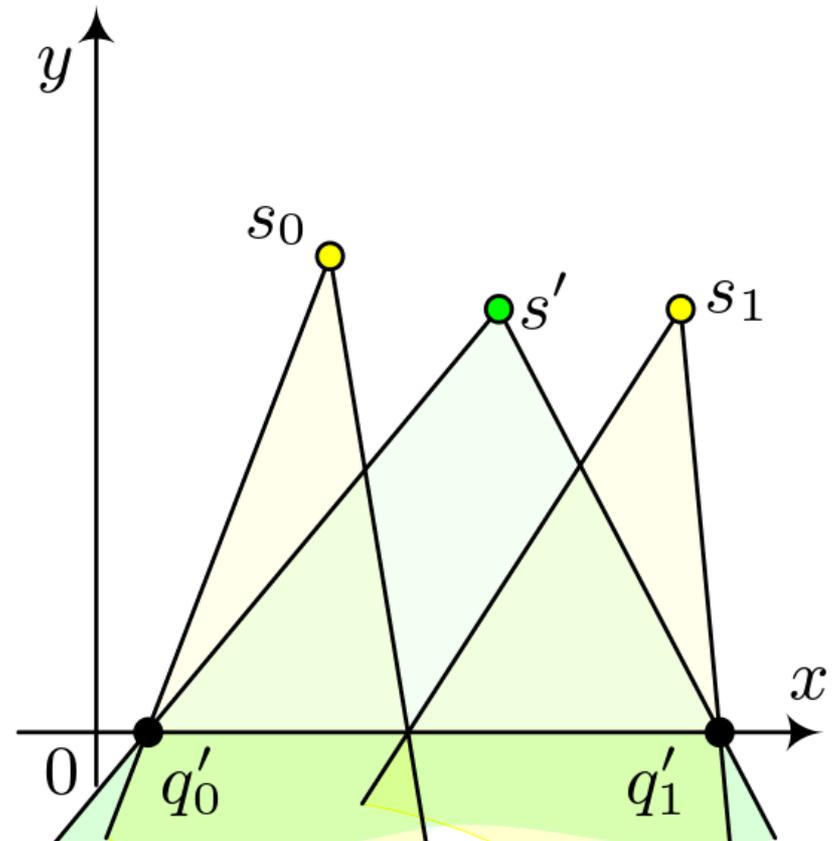


# Approximating Algorithm

- Basic idea: **merge windows**
  - Two original windows must have **direction** values  $\tau$  in agreement
  - Original windows must define **similar distances** on their union
  - Distance function along edge must be **continuous**
  - **Visibility** region of new window must cover the union of visibility regions of original windows

# Merging Windows

- Find a new source  $s = (x, y)$  and  $\sigma$ , for which the distances at the endpoints  $b_0$  and  $b_1$  are preserved:



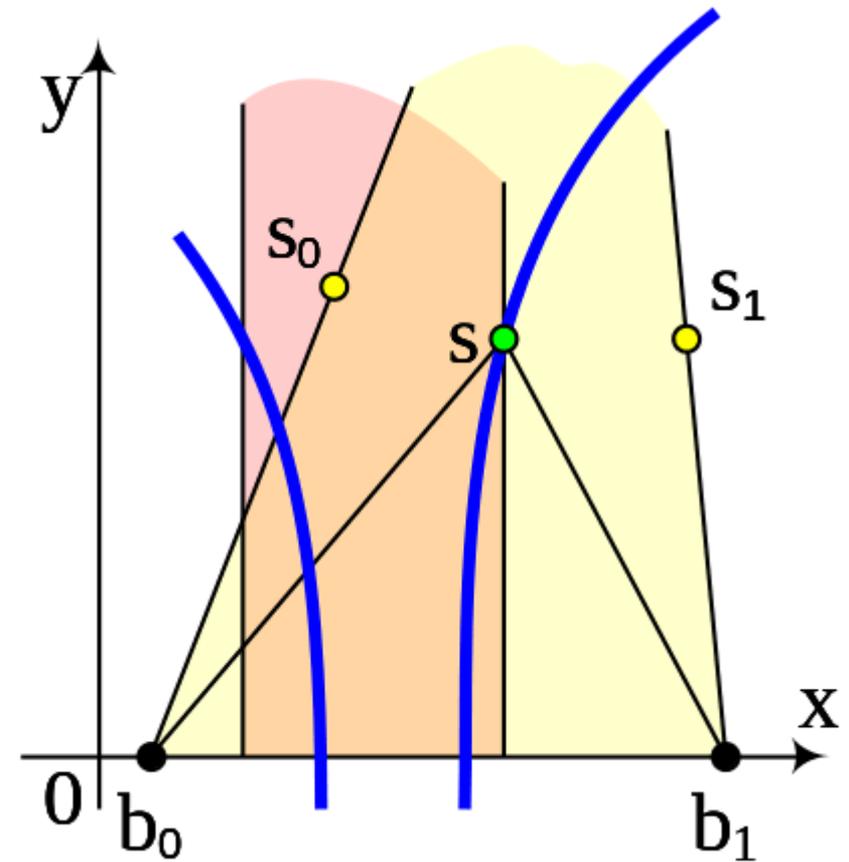
# Merging Windows

- Find a new source  $s = (x, y)$  and  $\sigma$ , for which the distances at the endpoints  $b_0$  and  $b_1$  are preserved:

$$(x - b_0)^2 + y^2 = (d_0 - \sigma)^2$$

$$(x - b_1)^2 + y^2 = (d_1 - \sigma)^2$$

- $s = (x, y)$  lies on a conic curve (quadratic algebraic curve)
- $s = (x, y)$  must lie in the yellow area - visibility must not be reduced) with  $y > 0$
- $\sigma > 0$  corresponds to the pink area



# Shortest path between two vertices

- Sequence of pruned searches to locate exact shortest path
- Exact algorithm is invoked only on a thin region surrounding geodesic
- Upper bound is the length of the approximate path obtained by Dijkstra search on edge graph, refined by output of approximation algorithm
- Lower bound initially represented by Euclidean distance, then replaced with output of approximation algorithm

