

# CENG 789 – Digital Geometry Processing

## 03- Point Sets

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# Point Sets

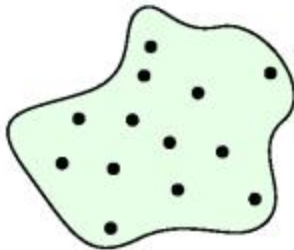
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- ✓ Focusing on point sets now.
  - ✓ Convex hulls, triangulations (e.g., Delaunay), Voronoi diagrams.
- ✓ Unlike polygonal vertices (prev lectures), point sets are unordered.

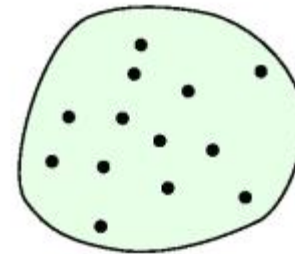
# Convexity

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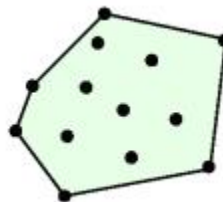
- ✓ Extend convex polygon idea to convex region.
- ✓ A region is convex if any 2 points  $a$  and  $b$  of the region are visible to each other, i.e., line segment  $ab$  lies within the region.
- ✓  $S$  = point set.



A nonconvex region enclosing  $S$



A convex region enclosing  $S$

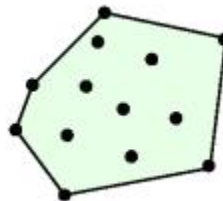


Convex hull of  $S$

# Convex Hull

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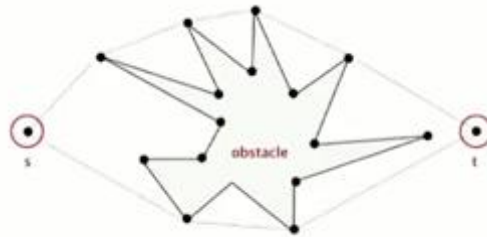
- ✓ Intuition: an elastic rubber stretched around all the nails/points.
- ✓ Motivation: useful for collision detection, robotics, statistics.
- ✓ Formal: smallest convex region containing the point set  $S$ .
- ✓ Formal: intersection of all convex regions containing  $S$ .
  - ✓ Cool but impractical definition to design a construction algo.
  - ✓ Practical definition based on visibility is better for algorithmic design.



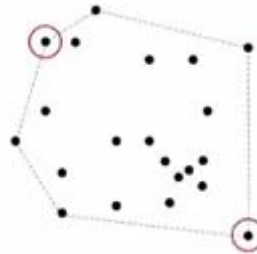
# Convex Hull

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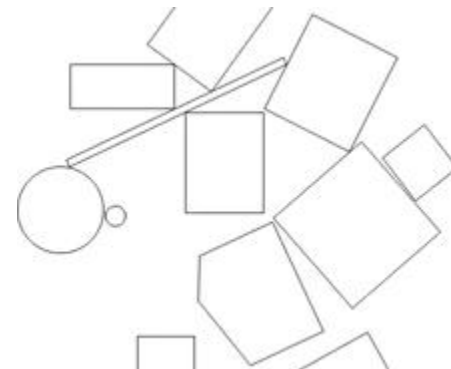
- ✓ Motivation: useful for collision detection, robotics, statistics.
- ✓ Find the shortest path from  $s$  to  $t$  that avoids the polygon obstacle.



- ✓ Find a pair of points w/ the largest Euclidean distance in between.



- ✓ More accurate collision detection.

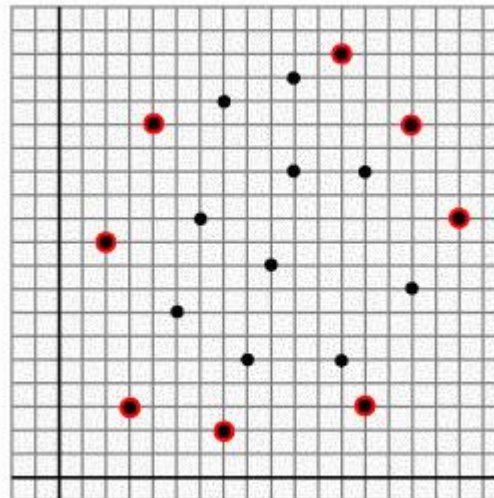


# Convex Hull Construction

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- ✓ Task: Identify hull vertices from the whole set of vertices.
- ✓ Easy for your eyes if given a paper w/ marked points.
- ✓ What if points are listed as coordinates (common case)?
  - ✓ Take the extreme points (min/max x/y) naturally.
  - ✓ What about the other hull points?
  - ✓ Identifying hull vertices w/o graphing is hard.
  - ✓ Even graphing/visualization fails for 1000+ points.

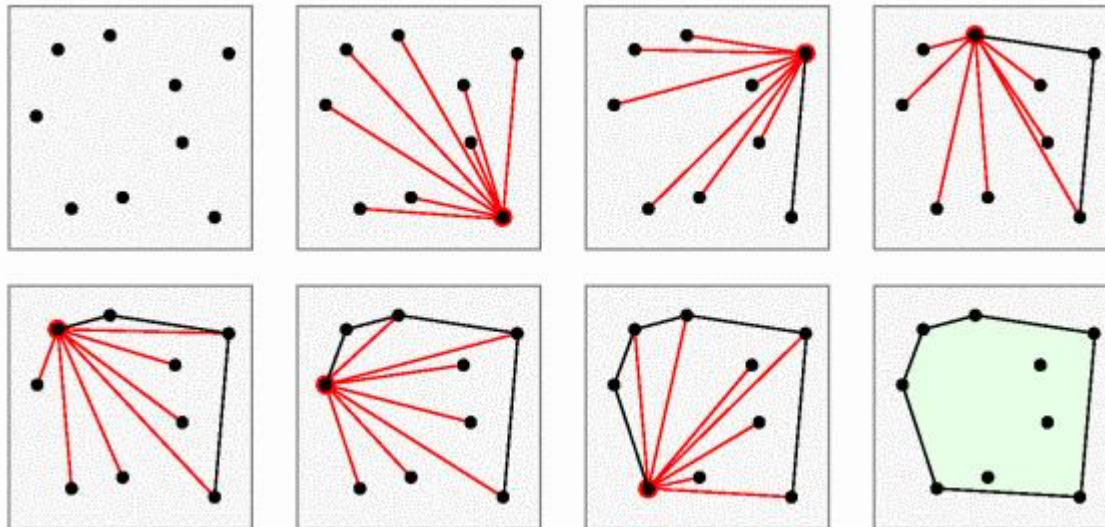
(2, 10)	(10, 17)
(3, 3)	(10, 13)
(4, 15)	(12, 5)
(5, 7)	(12, 18)
(6, 11)	(13, 3)
(7, 2)	(13, 13)
(7, 16)	(15, 8)
(8, 5)	(15, 15)
(9, 9)	(17, 11)



# Convex Hull Construction

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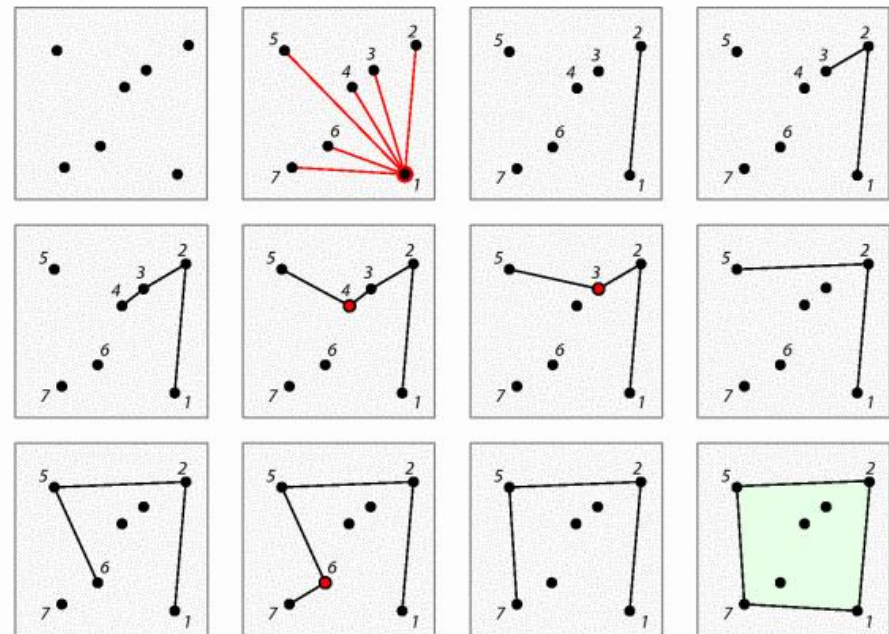
- ✓ Gift-wrapping algo to identify hull vertices in  $O(nh)$  time.
  - ✓  $n$  = # of points in point set  $S$ ;  $h$  = # of hull points.
  - ✓  $h$  could be as large as  $n$ , leading to  $O(n^2)$  ☹. Such a point set? Best set?
  - ✓ Extension to 3D exists.
- ✓ Start w/ the bottommost pnt. Draw a line segmnt to all others. Choose the next turn-pnt on our wrapping of  $S$  to be the pnt that makes the largest angle w/ the last constructed hull edge (initially with the horizontal line).



# Convex Hull Construction

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- ✓ Graham scan algo to identify hull vertices in  $O(n \log n)$  time.
  - ✓ No extension to 3D.
- ✓ Start w/ the bottommost pnt. Draw a line segmnt to all others. **Sort** the remaining points by the angles they make w/ the horizontal line. Construct the hull following this ordering (in triplets), adding pnts for left hull turns and deleting for right turns.

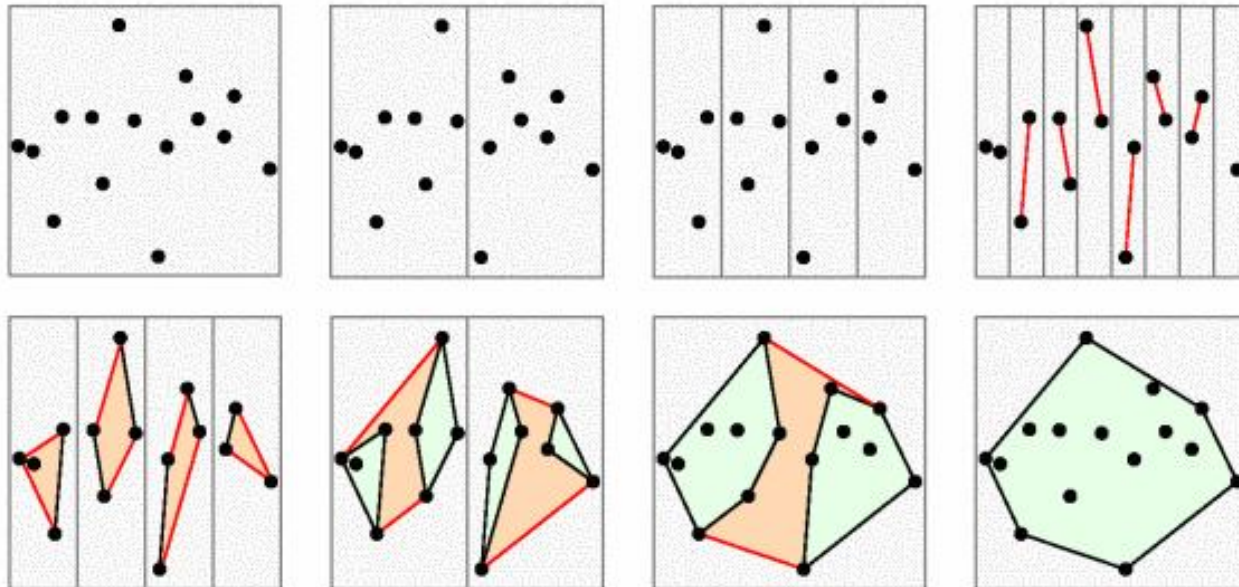




# Convex Hull Construction

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- ✓ Divide-and-conquer algo to identify hull vertices in  $O(n \log n)$  time.
  - ✓ Easy extension to 3D.
- ✓ Sort points by x-coordinate. Divide the points into 2 (nearly) equal groups. Compute the convex hull of each group recursively. Merge 2 groups with upper and lower supporting tangent lines.
- ✓ Recurse until 3 or less points in each group, whose conv hulls trivial.



# Convex Hull Construction

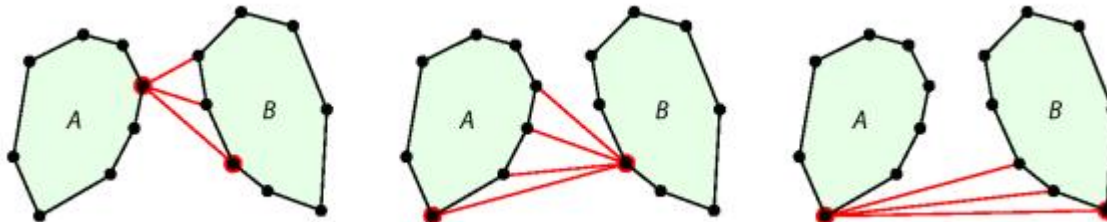
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- ✓ Divide-and-conquer algo to identify hull vertices in  $O(n \log n)$  time.
  - ✓ Easy extension to 3D.
- ✓ Merging is tricky. Not just highest/lowest points (circle vs. tall rectangle).

✓  $O(n^2)$ :



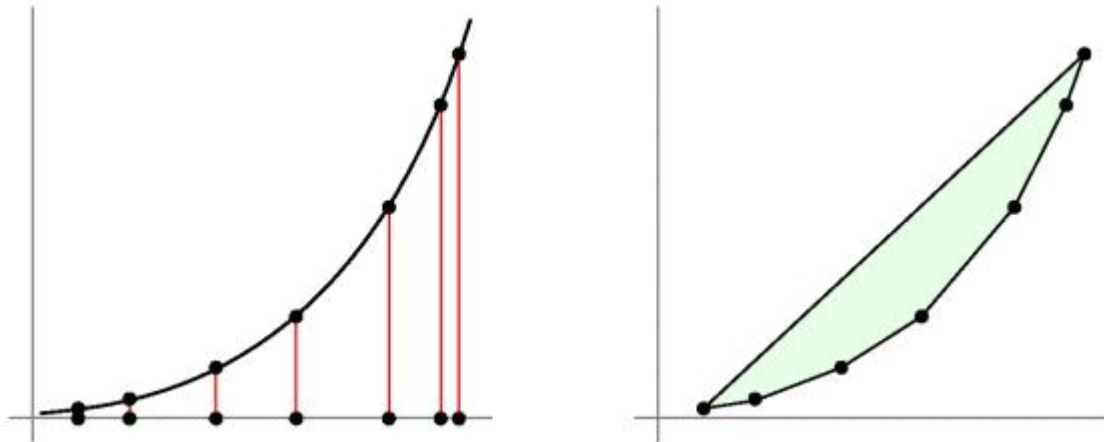
✓  $O(n)$ :



# Convex Hull ~ Sorting

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- ✓ Lower bound for any convex hull construction algo is  $O(n \log n)$  'cos Sorting problem reduces to Convex Hull problem:  $\text{Sorting} \leq_P \text{ConvHull}$ .
- ✓ Lower bound to sort  $n$  numbers is  $n \log n$  (decision-tree model).
- ✓ Given unsorted  $\{x_1, x_2, \dots, x_n\}$ , construct the set of points in the plane  $(x_i, x_i^2)$  as shown:

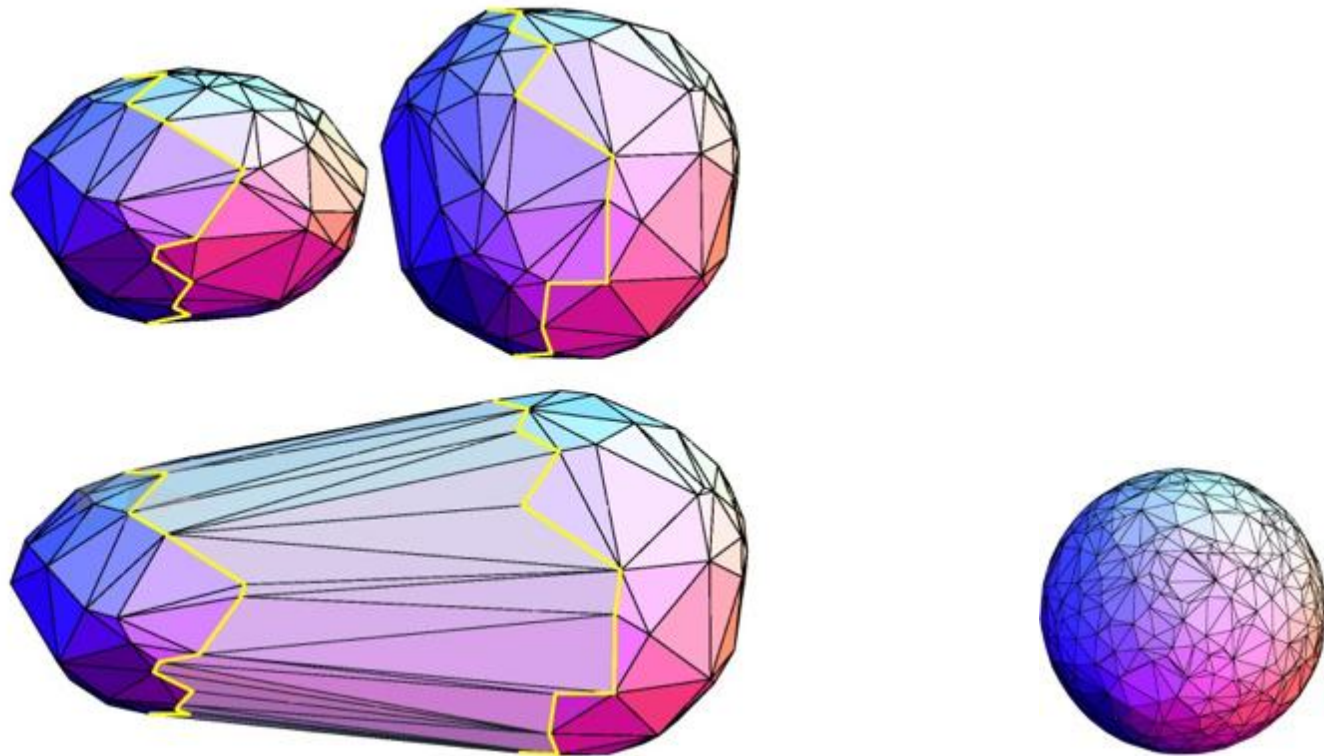


- ✓ Every point must be on the hull 'cos parabolas are convex.
- ✓ Convex hull of this parabola = identified ordered hull vertices = sorting.

# 3D Convex Hull Construction

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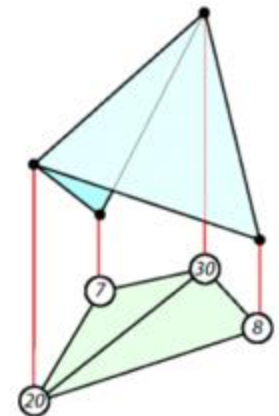
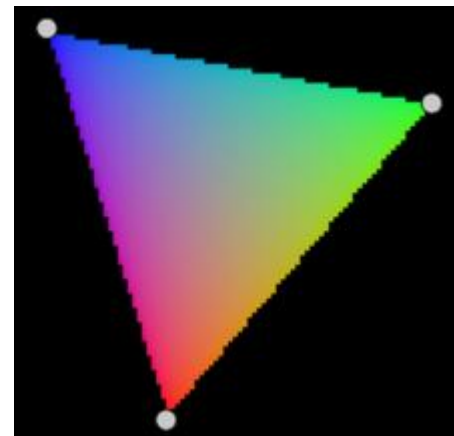
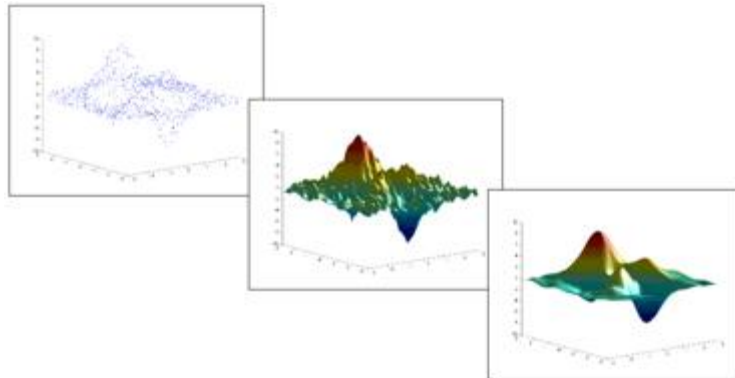
- ✓ Divide-and-conquer algo to identify hull vertices in  $O(n \log n)$  time.
- ✓ Sort points by x-coordinate. Divide the points into 2 (nearly) equal groups. Compute the convex hull of each group recursively. Merge 2 groups with upper and lower supporting tangent **planes**.



# Triangulations

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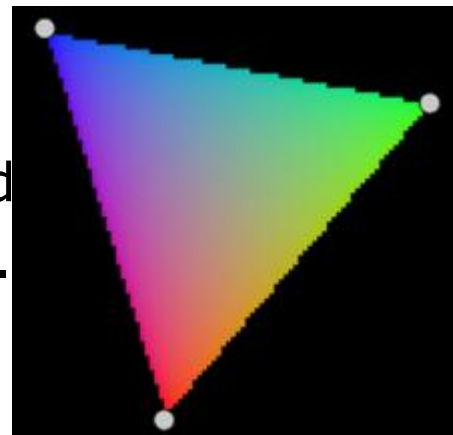
- ✓ Already did triangulation of a polygon.
- ✓ Let's do the triangulation of a structureless point set.
- ✓ Convex hull is about finding the *boundary* of a point set.
- ✓ Triangulation is about the *interior* of a point set.
- ✓ Motivation: put a structure to scattered data so that you can make interpolations at any point in space, e.g., interpolate color or height values on scattered data points.



# Triangulations

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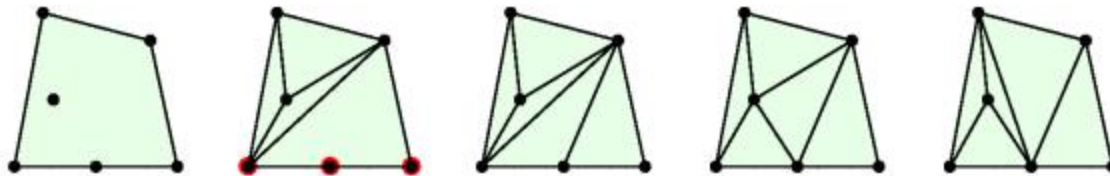
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- ✓ Triangulation is about the *interior* of a point set.
- ✓ Motivation: put a structure to scattered data so that you can make interpolations at any point in space, e.g., interpolate color or height values on scattered data points.
- ✓ Interpolation schemes are discussed in Deformation lecture slides 20-31.
- ✓ Focusing on triangulations here.



# Triangulations

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- ✓ With polygons we have boundary edges and internal diagonals.
- ✓ With point sets we just have edges to describe any line segment that includes 2 points of set as endpoints.
- ✓ Triangulation: of a point set  $S$  is subdivision of the plane by a maximal set of edges whose vertex set is  $S$ .

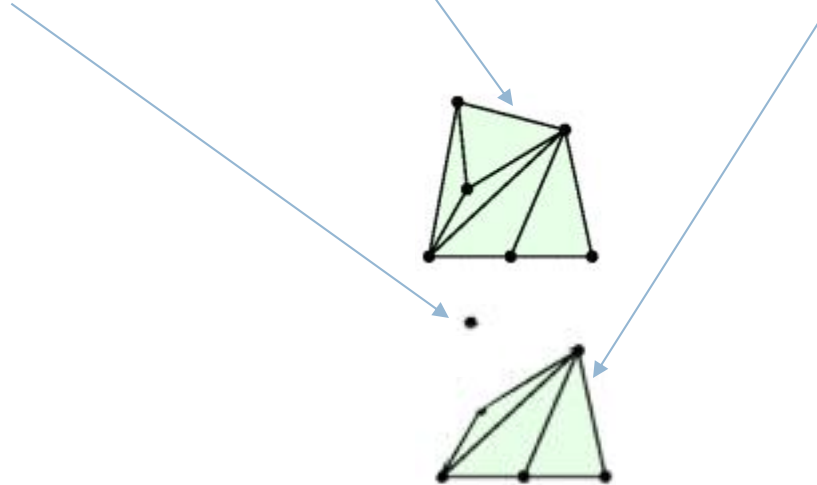


Convex hull   Subdivision   Triangulationssssssssss

# Triangulations

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- ✓ Prove that the edges of the convex hull will be in all triangulations.
- ✓ Proof: suppose not; an edge of the hull not in some triangulation. Then a point of  $S$  must be outside the convex hull, a contradiction.

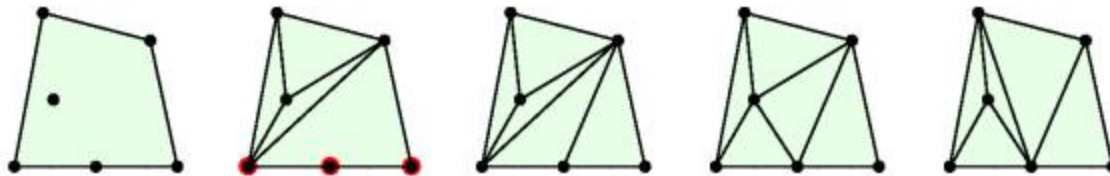




# Triangulations

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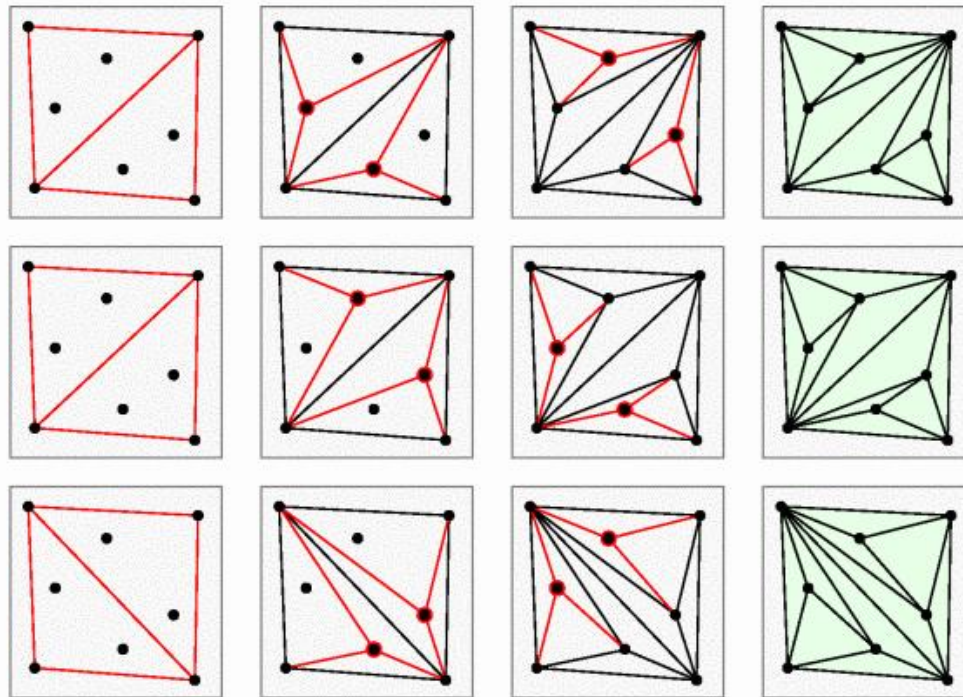
- ✓ Prove that all region of subdivision inside conv hull must be triangles.
- ✓ Proof: suppose not; some region has a region w/ 4+ edges. That region contradicts 'maximal set of edges' claim as it can be subdivided further.



# Triangulation Construction

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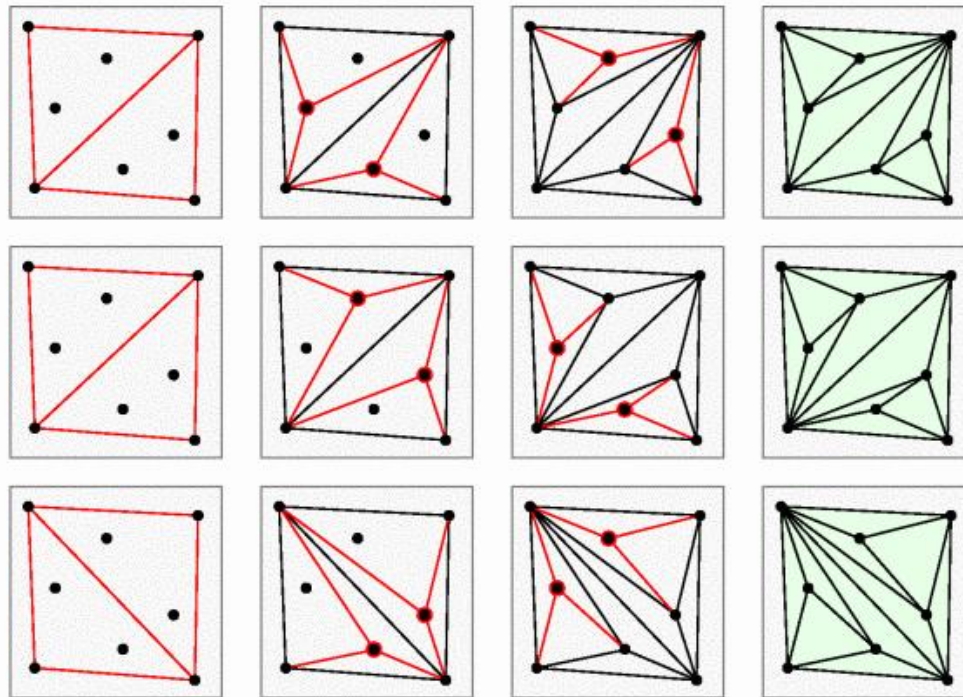
- ✓ A simple algorithm called triangle-splitting to triangulate pnt set  $S$ .
- ✓ Find convex hull of  $S$  and triangulate it as a polygon (previous lecture). Choose an interior point and draw edges to the 3 vertices of the triangle that contains it. Continue 'till all interior points are done.



# Triangulation Construction

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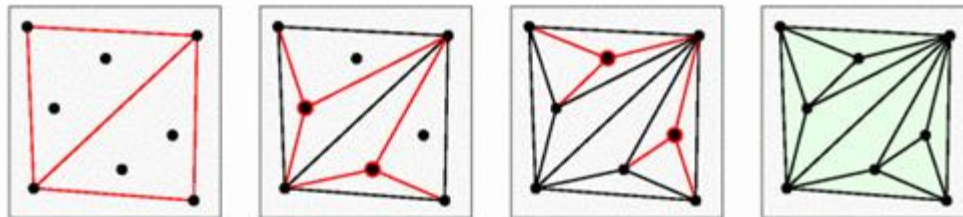
- ✓ A simple algorithm called triangle-splitting to triangulate pnt set  $S$ .
- ✓ Notice initial hull triangulation & processing order affect the output.



# Triangulation Construction

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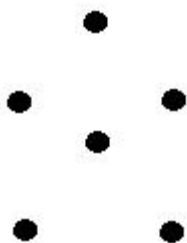
- ✓ A simple algorithm called triangle-splitting to triangulate pnt set  $S$ .
- ✓ Any triangulation of  $S$  using this algo has  $2k+h-2$  triangles, where  $k$  = # of interior points,  $h$  = # of hull points.
- ✓ Proof: triangulation of the convex hull, a polygon, has  $h-2$  triangles. An interior point replaces 1 triangle by 3, hence increasing the triangle count by 2.  $k$  interior pnts increase by  $2k$ ; hence  $h-2+2k$ .



# Triangulation Construction

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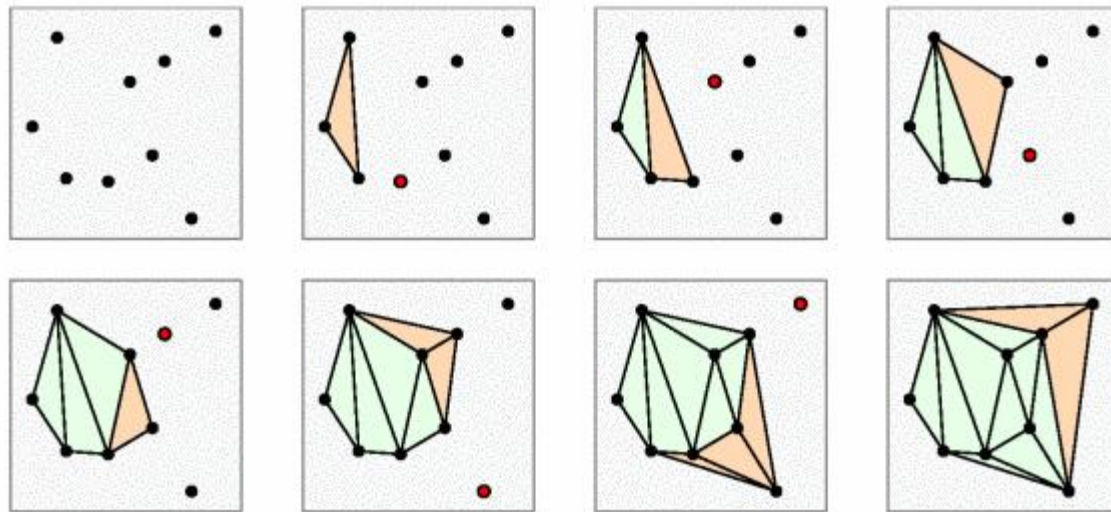
- ✓ A simple algorithm called triangle-splitting to triangulate pnt set  $S$ .
- ✓ Disprove that this algo produces all possible triangulations of  $S$ .
- ✓ Disproof: a counterexample that can't be produced by this algo.



# Triangulation Construction

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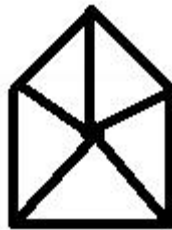
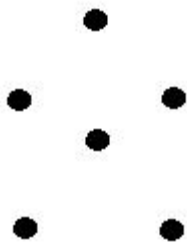
- ✓ Another algorithm called incremental to triangulate pnt set  $S$ .
- ✓ Sort the points of  $S$  by x-cords. First 3 points make a triangle. Consider the next point  $p$  in order and connect it w/ previously considered points which are visible to  $p$ . Continue incrementally.



# Triangulation Construction

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- ✓ Another algorithm called incremental to triangulate pnt set  $S$ .
- ✓ Disprove that this algo produces all possible triangulations of  $S$ .
- ✓ Disproof: a counterexample that can't be produced by this algo.



# Triangulation Construction

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- ✓ Any triangulation of  $S$  using any algo has  $2k+h-2$  triangles, where  $k = \#$  of interior points,  $h = \#$  of hull points,  $n = k+h = \#$  of points.
- ✓ Proof: using Euler's  $V - E + F = 2$  formula. Triangulation subdivides the plane into  $t+1$  faces,  $t$  triangles inside the hull, 1 big face outside the hull. Each triangle has 3 edges and outside face has  $h$  edges; hence  $3t+h$  edges. Each edge touches 2 faces  $\rightarrow 3t+h$  double counts edges  $\rightarrow$  there're indeed  $E = (3t+h)/2$  edges. Apply Euler:  
$$n - (3t+h)/2 + (t+1) = 2 \rightarrow t = 2n - h - 2 \rightarrow 2k + h - 2$$

bonus: use this  $t$  value and Euler to show  $\#$  edges =  $3k+2h-3$

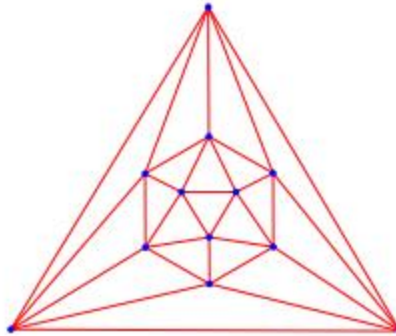
bonus: tetrahedralizations of same point set have different tets.



# Triangulations

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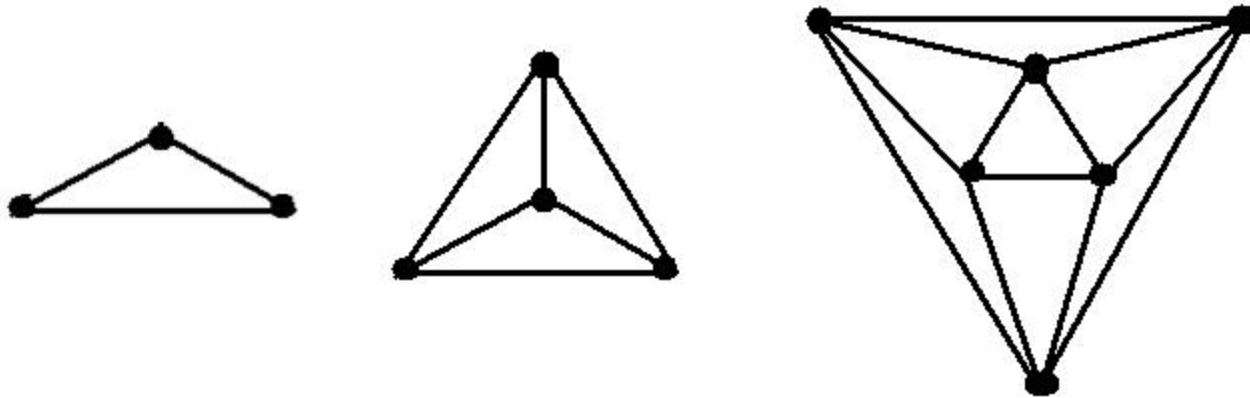
- ✓ What is interesting about this triangulation?



# Triangulations

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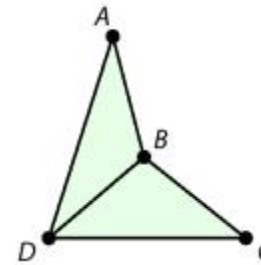
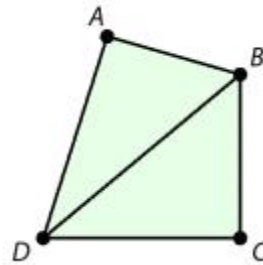
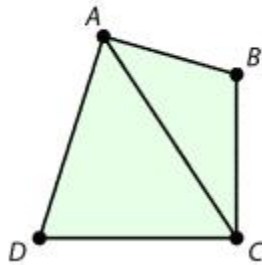
- ✓ What is interesting about these triangulations?



# Triangulations

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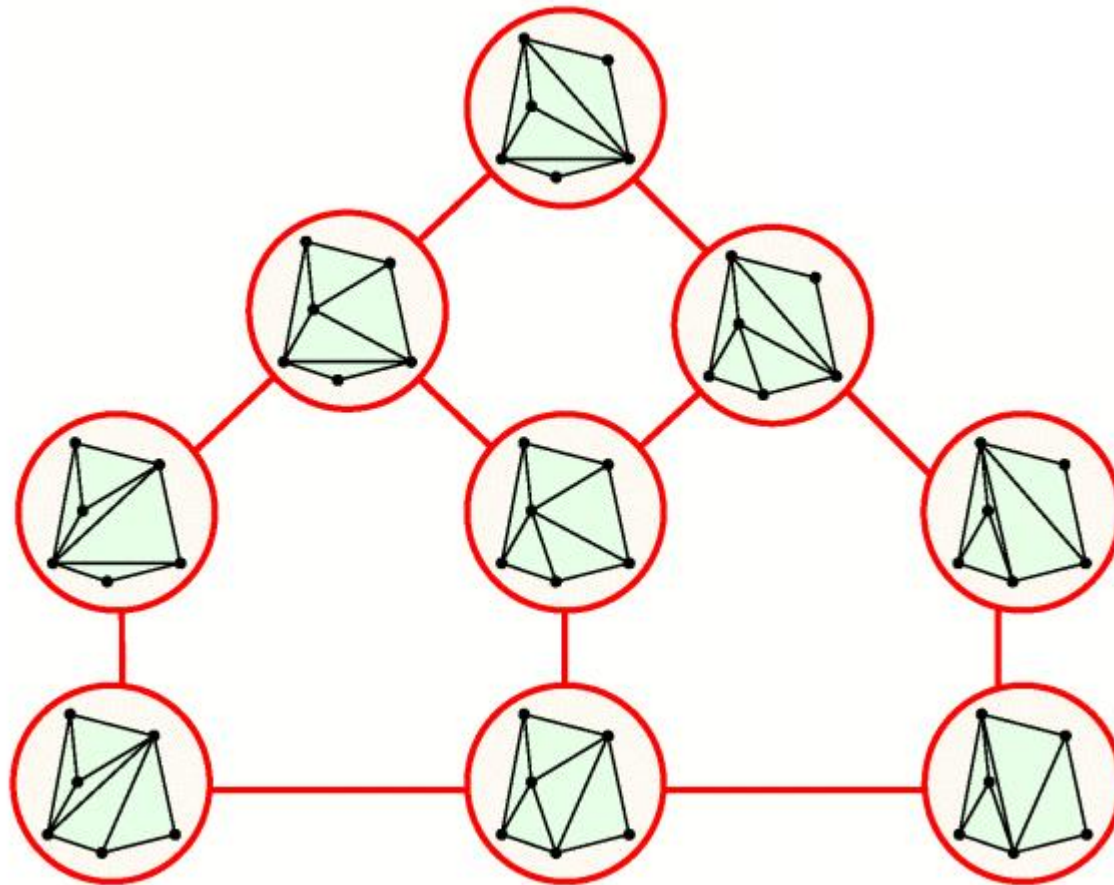
- ✓ Triangulations are related by the edge flip operation.



# Triangulations

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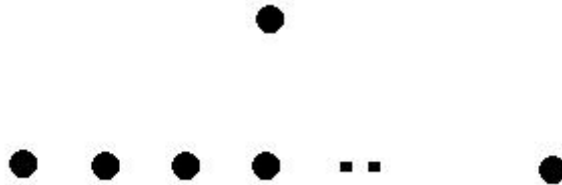
- ✓ Flip graph of a point set.
- ✓ Each node is a particular triangulation of the point set.



# Triangulations

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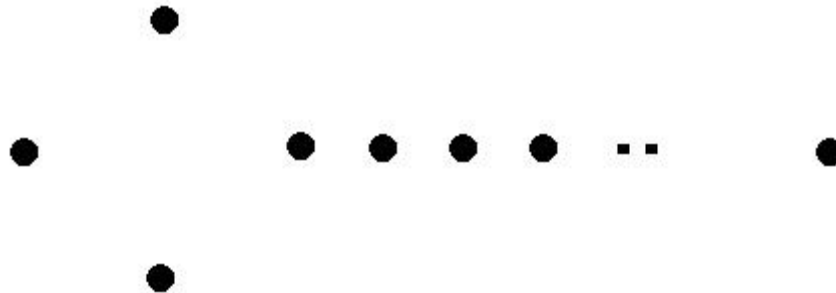
- ✓ For all  $n$  construct a point set with  $n$  points whose flip graph is a single node, i.e., no flip is possible.



# Triangulations

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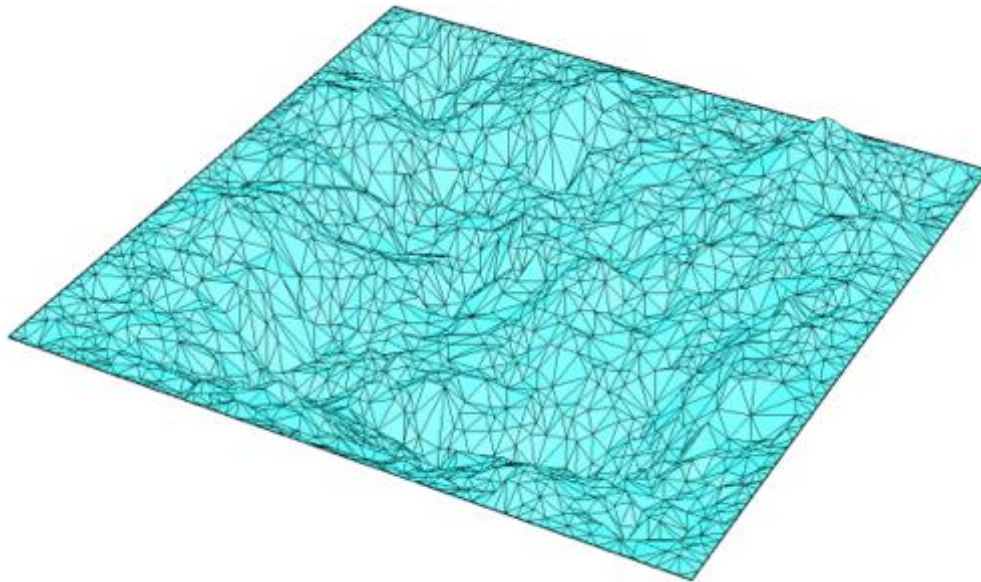
- ✓ For all  $n$  construct a point set with  $n$  points whose flip graph is 2 nodes, i.e., 2 flips are possible.



# Delaunay Triangulation

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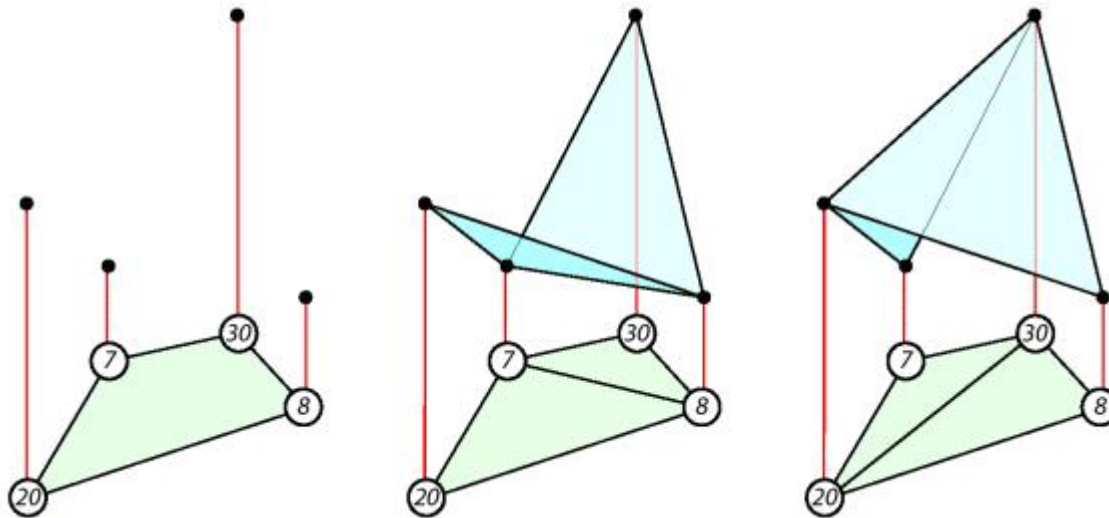
- ✓ 3d maps of earth's surface are constructed starting from a finite sample of points whose heights (altitude) are somehow measured.
- ✓ Triangulate these samples in 2D to approximate heights of nearby/unsampled points.
- ✓ Then lift each samples to its correct height, providing a piecewise-linear terrain of the earth.



# Delaunay Triangulation

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- ✓ Which triangulation is best for this reconstruction?



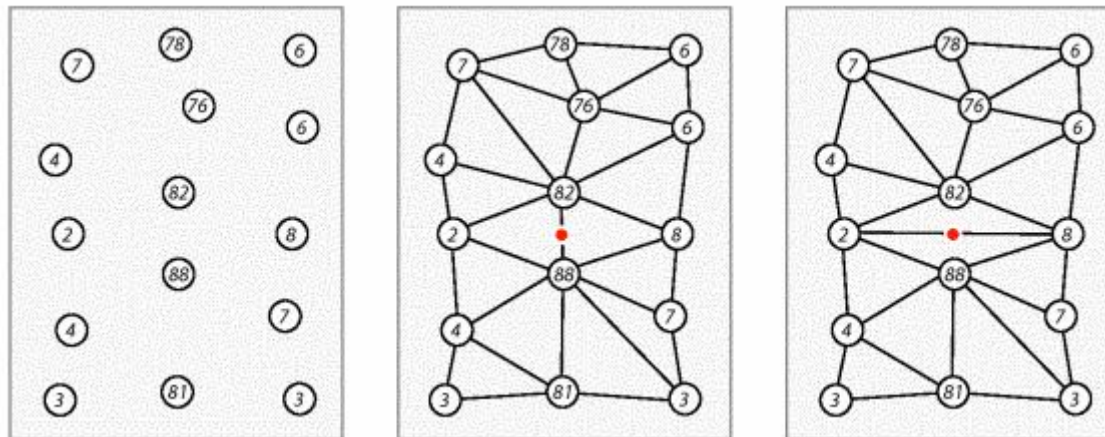
deep valley      steep mountain



# Delaunay Triangulation

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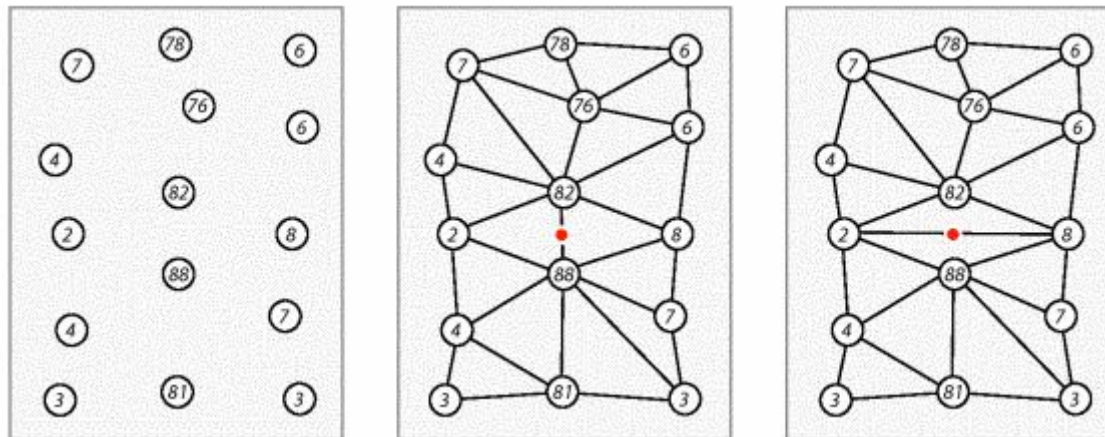
- ✓ Our intuition renders some terrains more natural than others.
- ✓ Left seems to be taken from a mountain; achieved by middle; but right part creates a deep valley that cuts the ridge in two.



# Delaunay Triangulation

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- ✓ Our intuition renders some terrains more natural than others.
- ✓ What is cool about the middle one?
  - ✓ No skinny triangles compared to the right one.



- ✓ Avoid skinny triangles by picking a triangulation that maximizes the minimum angle: Delaunay Triangulation.

# Delaunay Triangulation

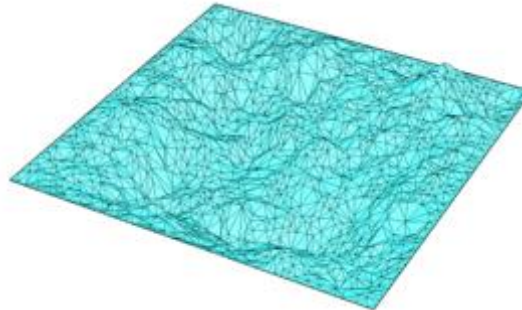
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- ✓ Angle sequence  $(a_1, a_2, \dots, a_{3n})$  for a triangulation w/  $n$  triangles, sorted from smallest  $a_1$  to largest  $a_{3n}$ .
- ✓ Using the angle sequence we can compare 2 different triangulations.
- ✓ Number of triangles is constant  $\rightarrow$  angle sequences have same length.
  
- ✓ T1: (20, 30, 45, 65, 70, 130)
- ✓ T2: (20, 30, 45, 60, 75, 130)
  - ✓ T1 > T2 (fatter) since  $65 > 60$ .
  - ✓ The fattest triangulation is the one we desire.
  - ✓ Find it easily by edge flips.

# Delaunay Triangulation

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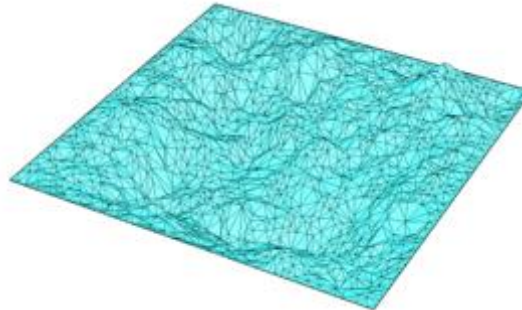
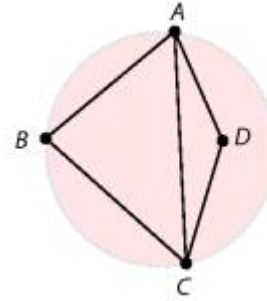
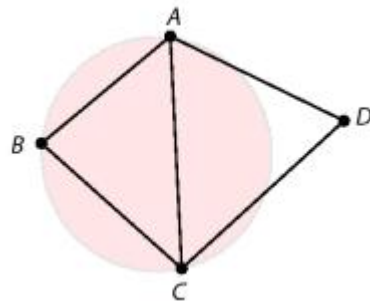
- ✓ Let  $e$  be edge of a triangulation  $T_1$  and let  $Q$  be the quad formed by 2 triangles having  $e$  as common edge. If  $Q$  is convex, let  $T_2$  be the triangulation after flipping  $e$  in  $T_1$ . We say  $e$  is legal if  $T_1 \geq T_2$  and illegal otherwise.
- ✓ Convex hull edges of a triangulation are also legal.
- ✓ Delaunay triangulation of a point set is a triangulation that only has legal edges.
- ✓ Delaunay triangulation is a triangulation in which interior of any circumcircle of a triangle is empty, i.e., no points inside circumcircle.



# Delaunay Triangulation

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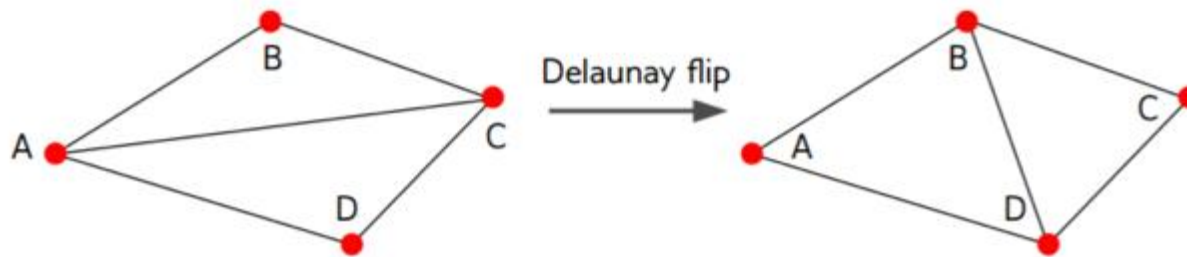
- ✓ Put simply, a triangulation is Delaunay if no point lies within the circumcircle of any triangle.
  - ✓ It also happens to maximize the minimum angle of any triangle, which is why it's useful.



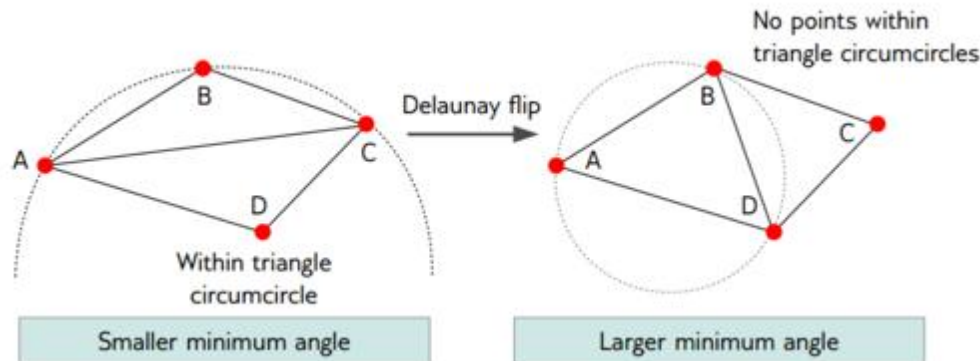
# Delaunay Triangulation Construction

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- ✓ Start with any triangulation  $T$  (e.g., slide 17 or 21). If  $T$  has an illegal edge, flip the edge and make it legal. Continue flipping illegal edges.



- ✓ Put simply, flip edge if the minimum of the 6 angles increases (above).
  - ✓ Guaranteed to converge (since minimum angle increases).

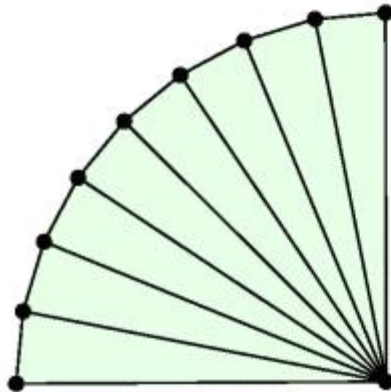


- ✓ Only 1 of the 2 configurations satisfies the circumcircle property.

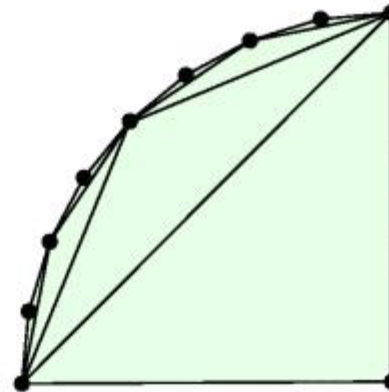
# Other Special Triangulations

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- ✓ Minimum weight triangulation (MWT).
- ✓ Min ink to draw it compared to all other triangulations.
- ✓ App: Minimize wiring cost of building a network.
- ✓ Skinny triangles have long edges so use Delaunay to avoid them?
- ✓ Not really.



Delaunay

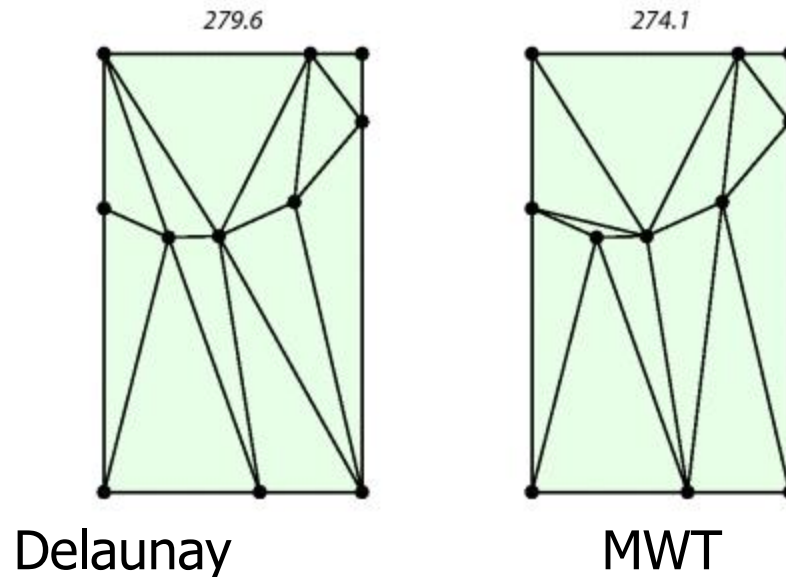


MWT

# Other Special Triangulations

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- ✓ Skinny triangles have long edges so use Delaunay to avoid them?
- ✓ Not really.



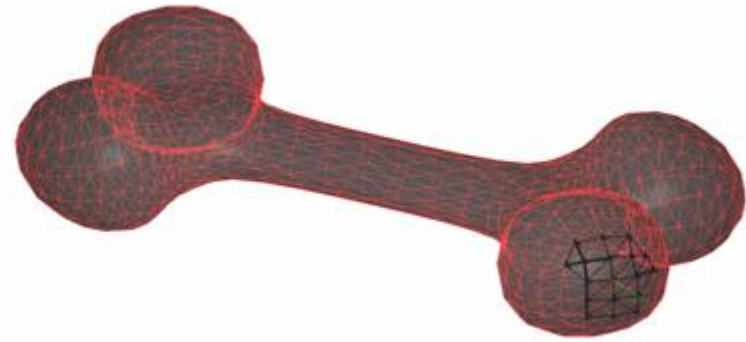


# Other Special Triangulations

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- ✓ Minimum weight triangulation.
- ✓ Proved to be NP-hard: no polynomial-time algo known.
- ✓ Alternative: instead of complete triangulation of a point set, use a tree that spans the point set; minimum spanning tree (MST).
- ✓ Theorem: MST of a point set is a subset of its Delaunay triangulation.
  - ✓ Justifies the intuition that Delaunay triangulation is weight-minimizing in some sense.

# Triangulations of 3D Points



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- ✓ Find local neighborhood  $L_i$  of each point  $p_i$  in the 3D point cloud input.
  - ✓ Closest  $k$  points (using a k-d tree).
- ✓ For each  $L_i$  compute tangent plane using PCA.
  - ✓ If the input is oriented, just take the average of all normals for the plane normal. Use the mean pnt as the plane pnt (for both oriented/unoriented).
- ✓ Project all points in  $L_i$  to the tangent plane and compute their 2D Delaunay triangulation  $D_i$ .
- ✓  $D_i$  is a set of edges:  $D_i = \{e_i^1, e_i^2, \dots, e_i^{noe(i)}\}$  where  $noe(i)$  is the number of edges of the  $i^{\text{th}}$  Delaunay triangulation.
- ✓ Final triangulation is the composition of all  $N$  local triangulations:

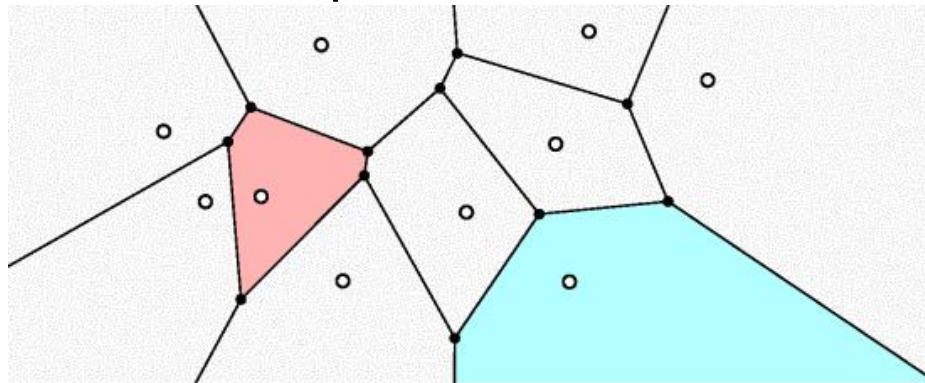
$$D = \bigcup_1^N \{e_i^1, e_i^2, \dots, e_i^{noe(i)}\}.$$

- ✓ Note that global  $D$  not necessarily a 2-manifold. Set  $k = 0.02n$  and restrict value to  $[8, 12]$ .
- ✓ See the Crust Algorithm in Surface Reconstruction slides as well (more principled).

# Voronoi Diagrams

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- ✓ Convex hull  $\sim$  boundary of  $S$ .
- ✓ Triangulation  $\sim$  interior of  $S$ .
- ✓ Voronoi  $\sim$  points not in  $S$ .
- ✓ Set of points  $x$  not in  $S$ , that are at least as close to Voronoi site  $p$  as to any other Voronoi site  $q$  in  $S$ .

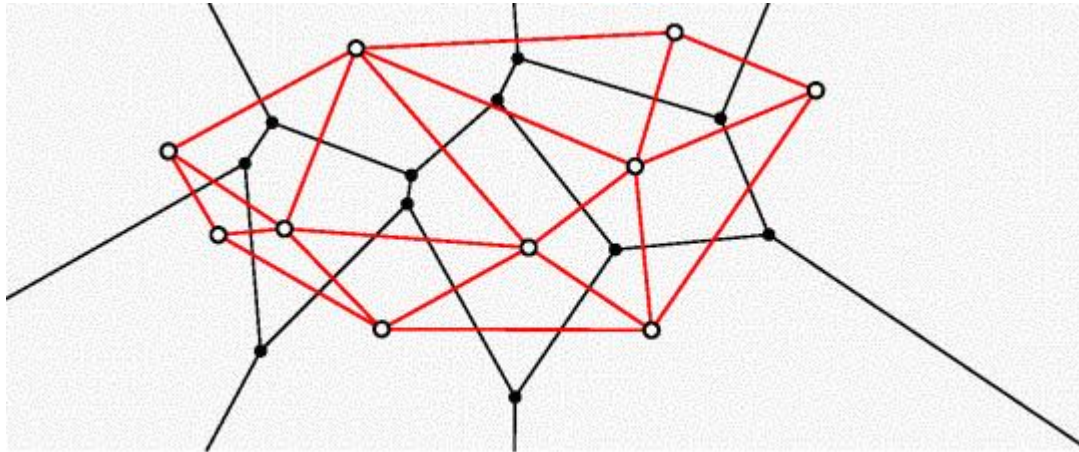


- ✓ Voronoi region for the site  $p_k$  is  $R_k = \{x \mid ||x - p_k|| \leq ||x - p_j||, j \neq k\}$ .
- ✓ Voronoi vertices (black dots) are equidistant from 3 neighboring sites.
- ✓ [https://youtu.be/yDMtGT0b\\_kg](https://youtu.be/yDMtGT0b_kg)

# Voronoi Diagrams Application # 1

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- ✓ Convex hull  $\sim$  boundary of  $S$ .
- ✓ Triangulation  $\sim$  interior of  $S$ .
- ✓ Voronoi  $\sim$  points not in  $S$ .
- ✓ Dual of Delaunay triangulation: If Voronoi regions adjacent, connect their sites w/ a Delaunay edge.

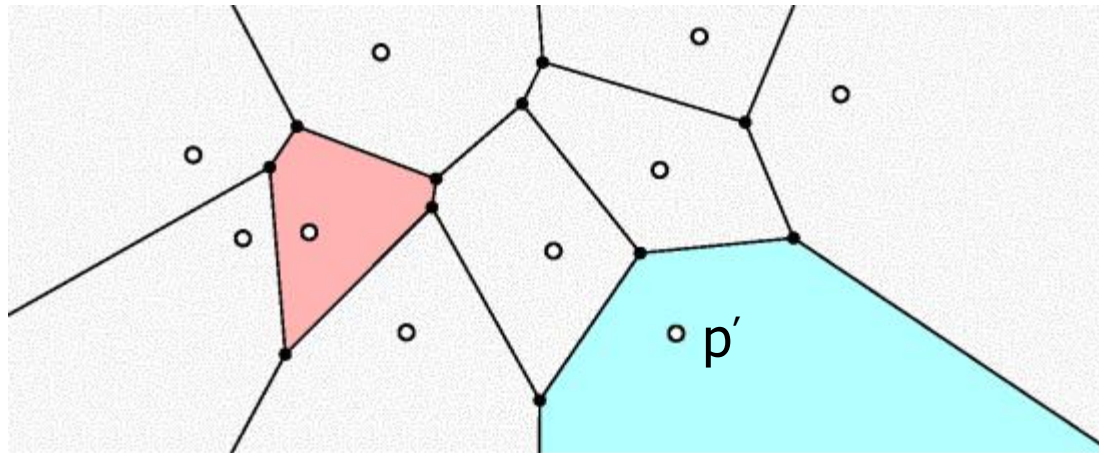


- ✓ <https://youtu.be/PWlo7PvtQVw?t=2154>

# Voronoi Diagrams Application # 2

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- ✓ Convex hull  $\sim$  boundary of  $S$ .
- ✓ Triangulation  $\sim$  interior of  $S$ .
- ✓ Voronoi  $\sim$  points not in  $S$ .
- ✓ Set of points  $x$  not in  $S$ , that are at least as close to Voronoi site  $p$  as to any other Voronoi site  $q$  in  $S$ .

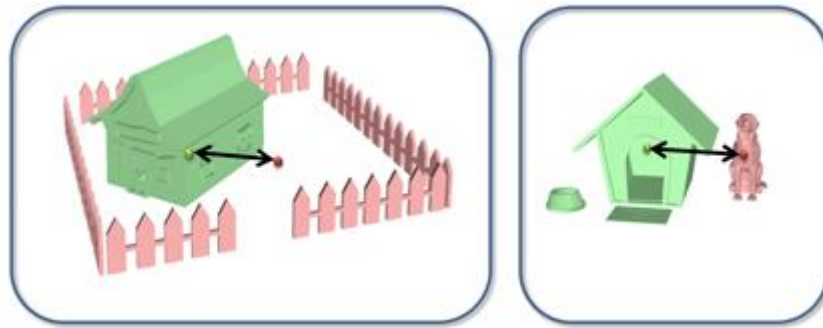


- ✓ Nearest service handling: e.g., post office – apartment assignment. All the cyan apartments will be served from the site  $p'$ .

# Voronoi Diagrams Application # 3

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- ✓ Another application of Voronoi Diagrams: Interaction Representation.
  - ✓ Also related to finding the separating hyperplane w/ max margin in the SVM.
- ✓ Distance-based representation (not so powerful).



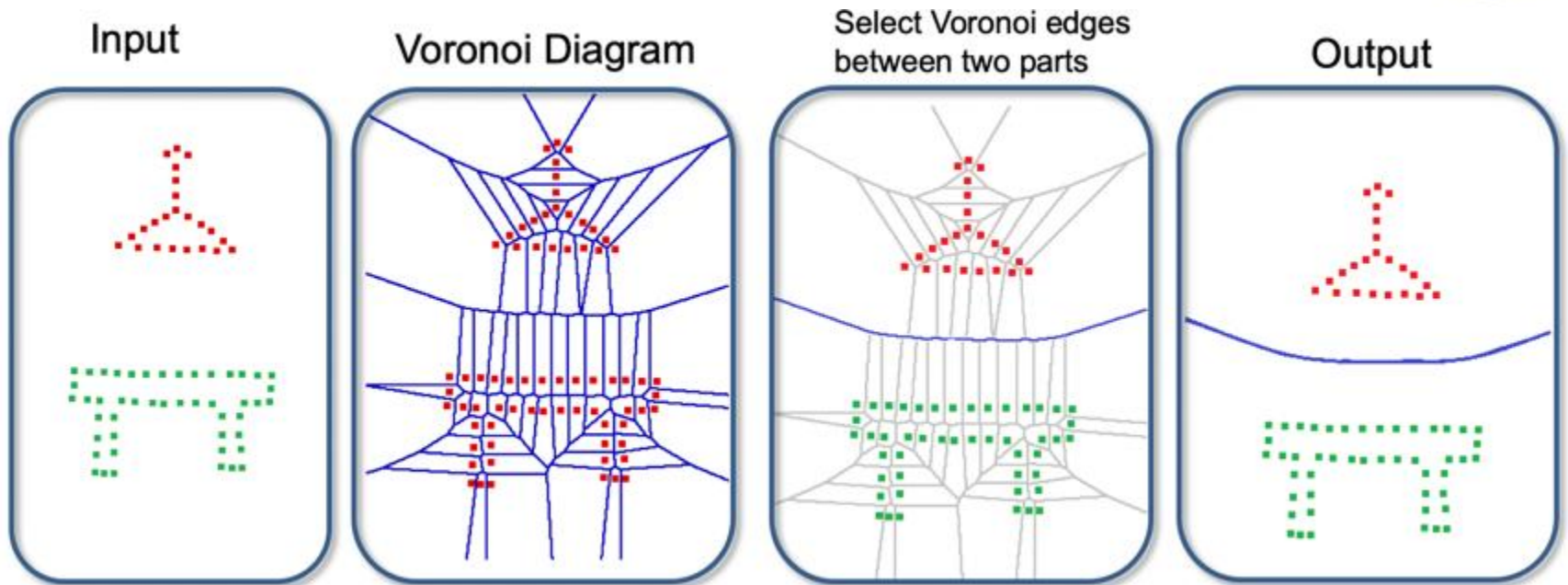
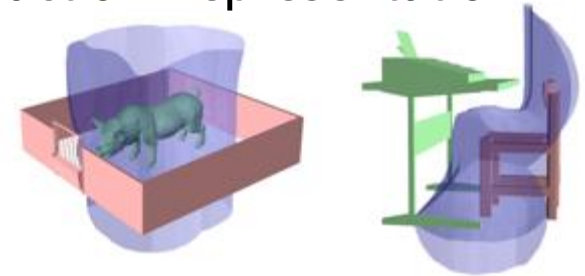
- ✓ Interaction bisector representation (powerful).



# Voronoi Diagrams Application # 3

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- ✓ Another application of Voronoi Diagrams: Interaction Representation.
  - ✓ Also related to the hyperplanes in SVM.
- ✓ Interaction bisector representation (powerful).





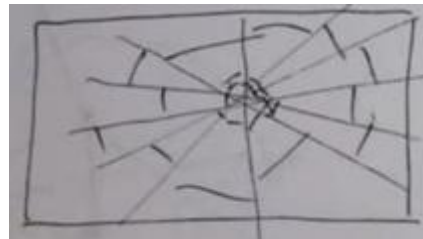
# Voronoi Diagrams Application # 4

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- ✓ Another application: Pre-fracture pattern creation for tearing/breaking.



- ✓ If you know where the bullet will hit in advance, pre-fracture possible.
  - ✓ Distribute dense Voronoi sites around bullet point, sparse everywhere else.



- ✓ Fracture in action: <https://youtu.be/Lg2dqFCU67Q>



# Voronoi Diagrams Application # 4

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- ✓ Another application: Pre-fracture pattern creation for tearing/breaking.



- ✓ If you don't know the contact point, rotate on the fly to match the dense Voronoi part with the collision point during simulation, e.g., for sphere.

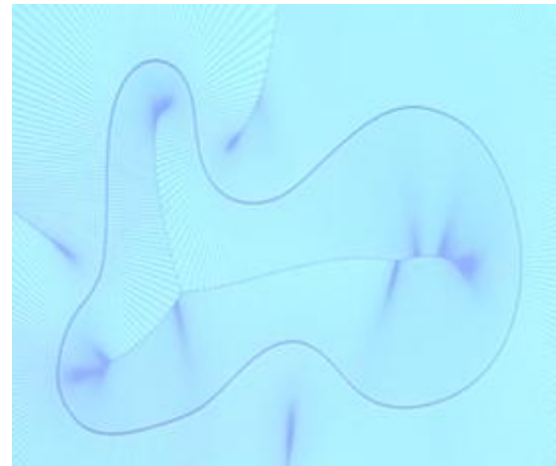
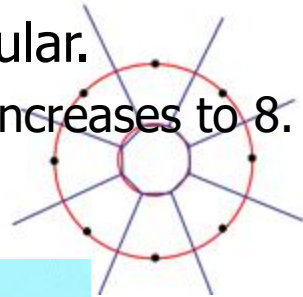


- ✓ For non-symmetric shapes, a good research problem: Real Time Dynamic Fracture with Volumetric Approximate Convex Decompositions.

# Voronoi Diagrams Application # 5

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- ✓ Another application: Medial axis extraction.
- ✓ Medial axis is the set of interior points that have more than 1 closest boundary points.
  - ✓ Superset of curve skeleton.
  - ✓ Useful in many apps, e.g., deform skeleton and boundary follows (LBS).
- ✓ Each Voronoi vertex has degree 3 if no 4 points are co-circular.
  - ✓ Smaller circle shrinks to nothing (no site inside) and degree increases to 8.
- ✓ Compute Voronoi diagram of the (boundary) points.
- ✓ Voronoi vertices give you the med axis:



# Voronoi Diagrams Application # 5

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- ✓ Another application: Medial axis extraction.
- ✓ Medial axis is the set of interior points that have more than 1 closest boundary points.
  - ✓ Superset of curve skeleton.
  - ✓ Useful in many apps, e.g., deform skeleton and boundary follows (LBS).
- ✓ Each Voronoi vertex has degree 3 if no 4 points are co-circular.
  - ✓ Smaller circle shrinks to nothing (no site inside) and degree increases to 8.
- ✓ Compute Voronoi diagram of the (boundary) points.
- ✓ Sensitive to noise, 1 input pnt inside boundary dooms the medial axis:

