

26<sup>th</sup> Symposium on Computational Geometry

# Dynamic Well-Spaced Point Sets

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June 4, 2011

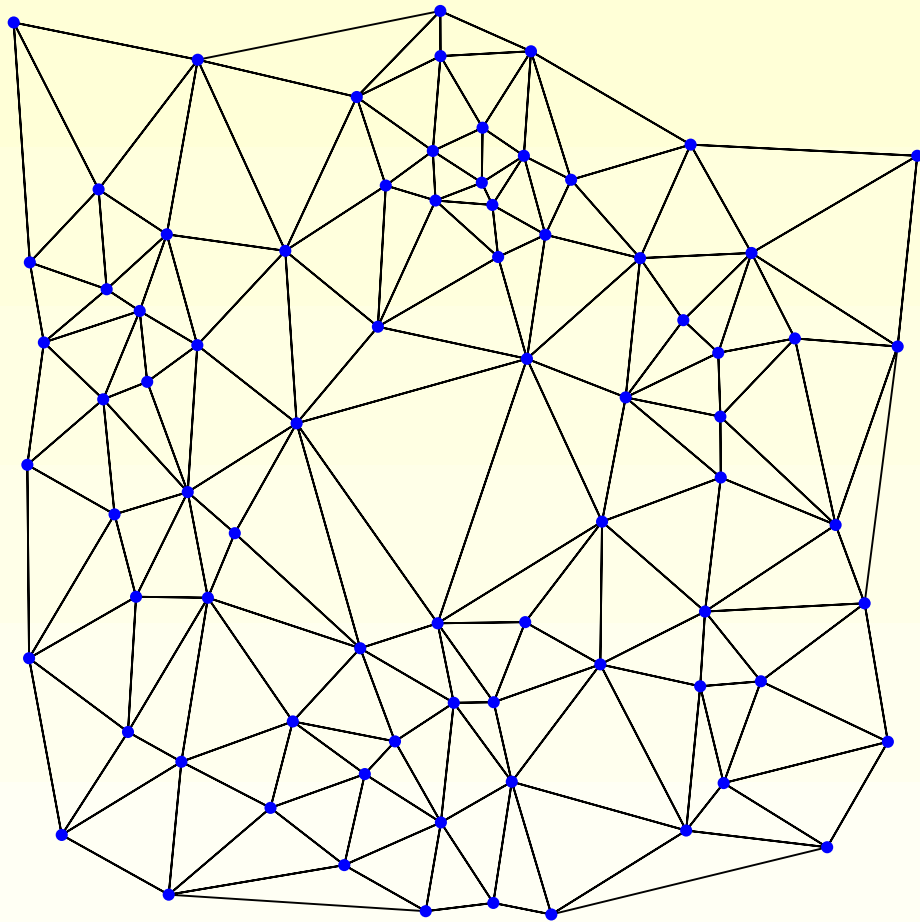
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<sup>a</sup>Max Planck Institute (SWS)

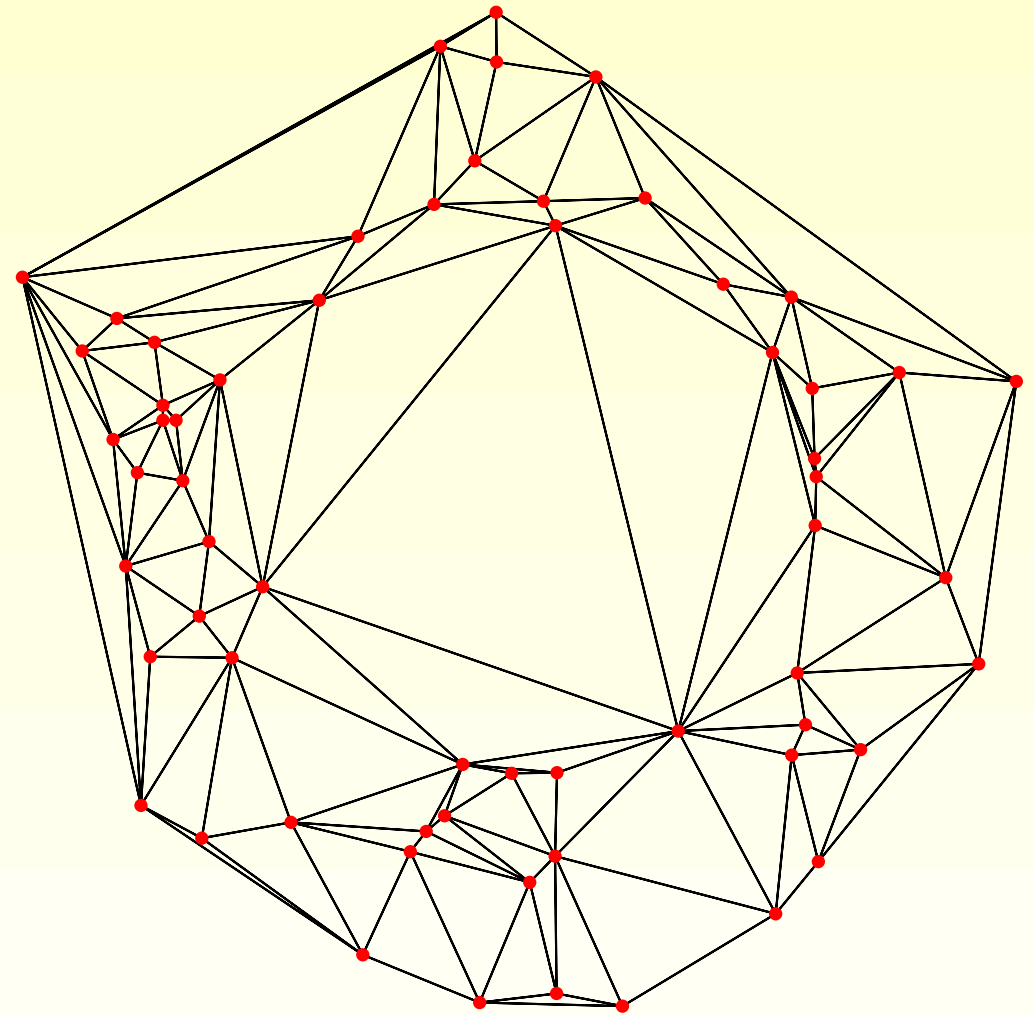
<sup>b</sup>Toyota Technological Institute at Chicago

<sup>c</sup>University of Chicago

# Delaunay Triangulations, 2D Examples



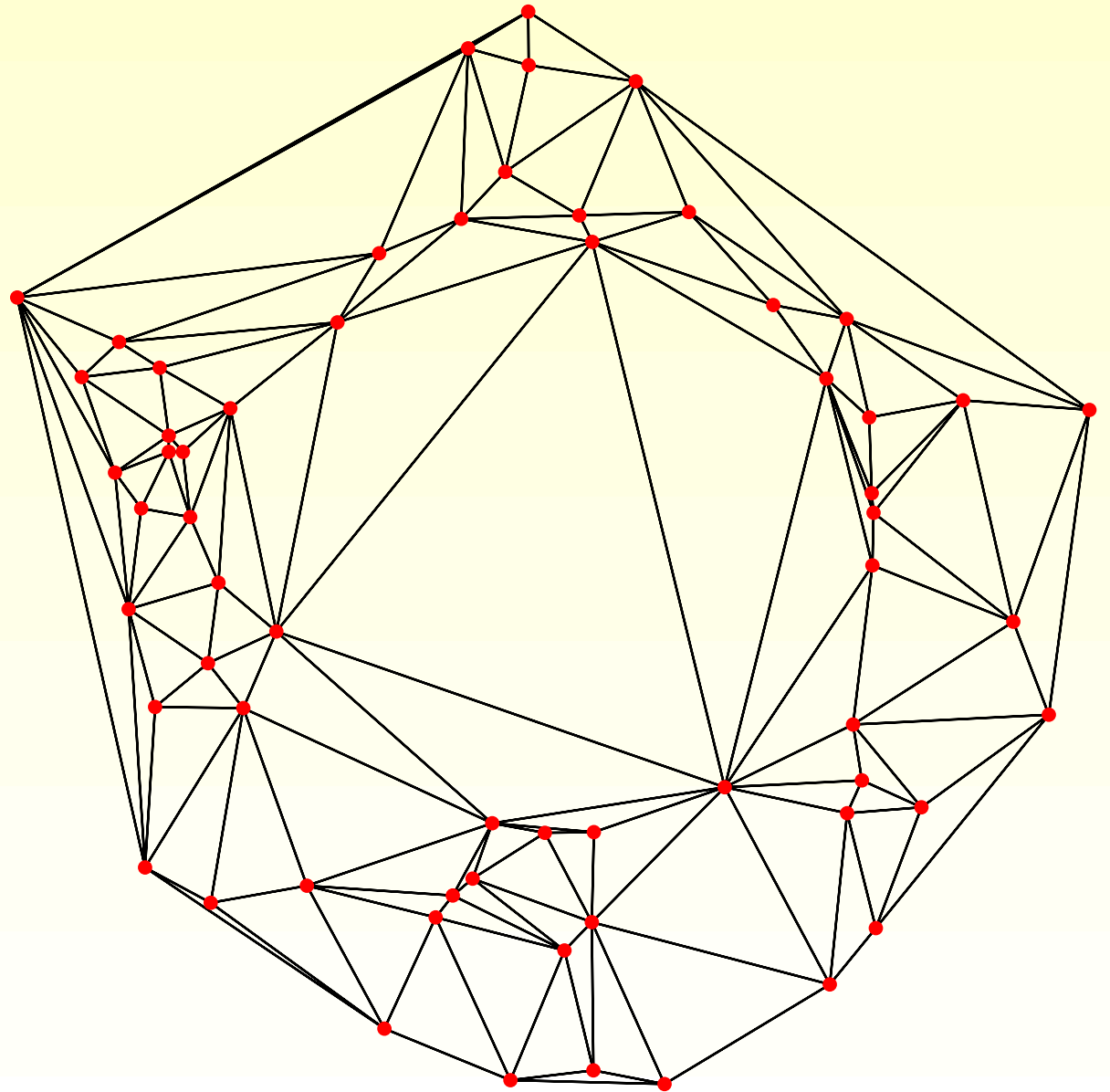
No small angles



Skinny triangles

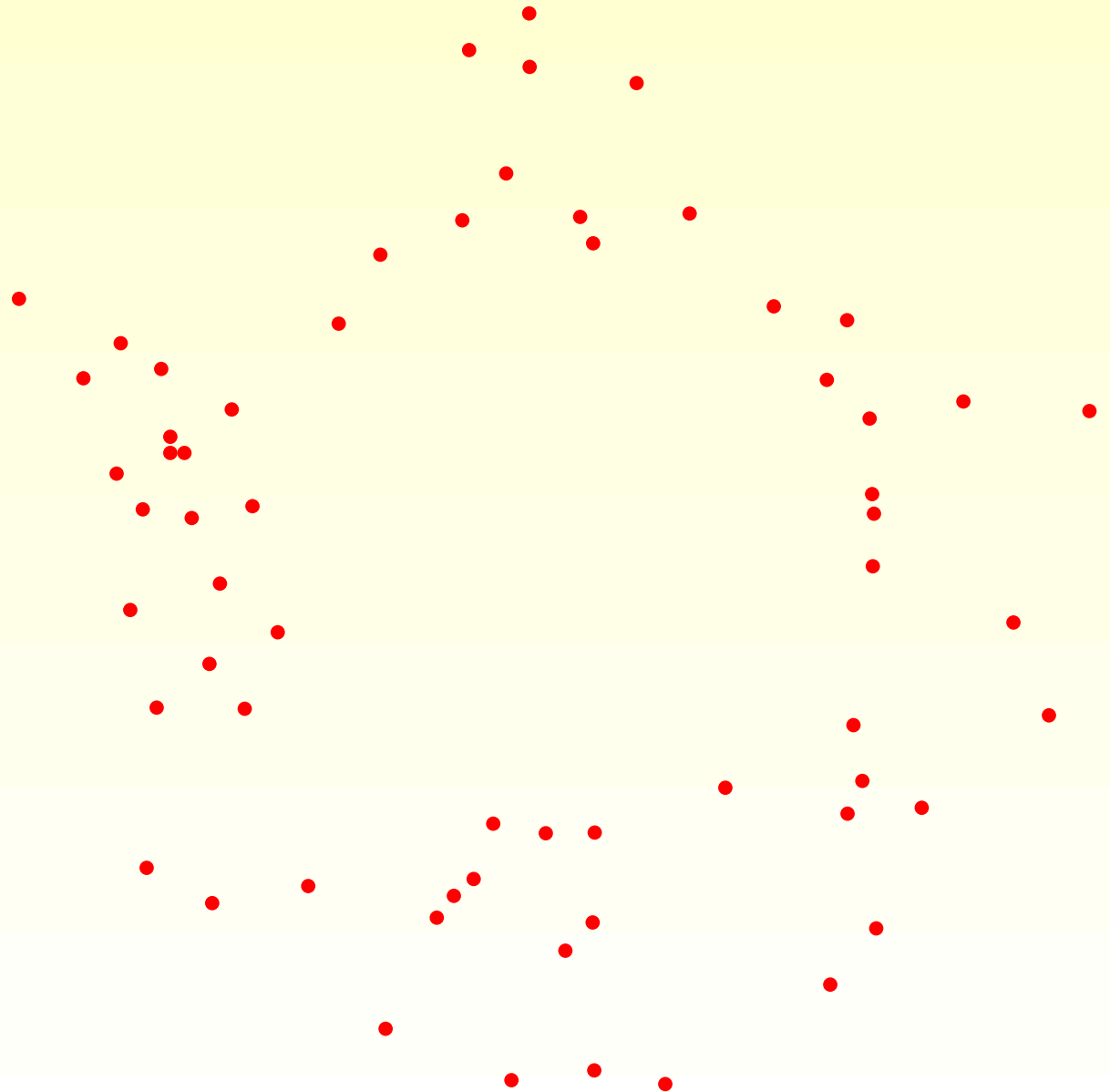
# Mesh Refinement

**Input** set of points  
does not have a  
good triangulation



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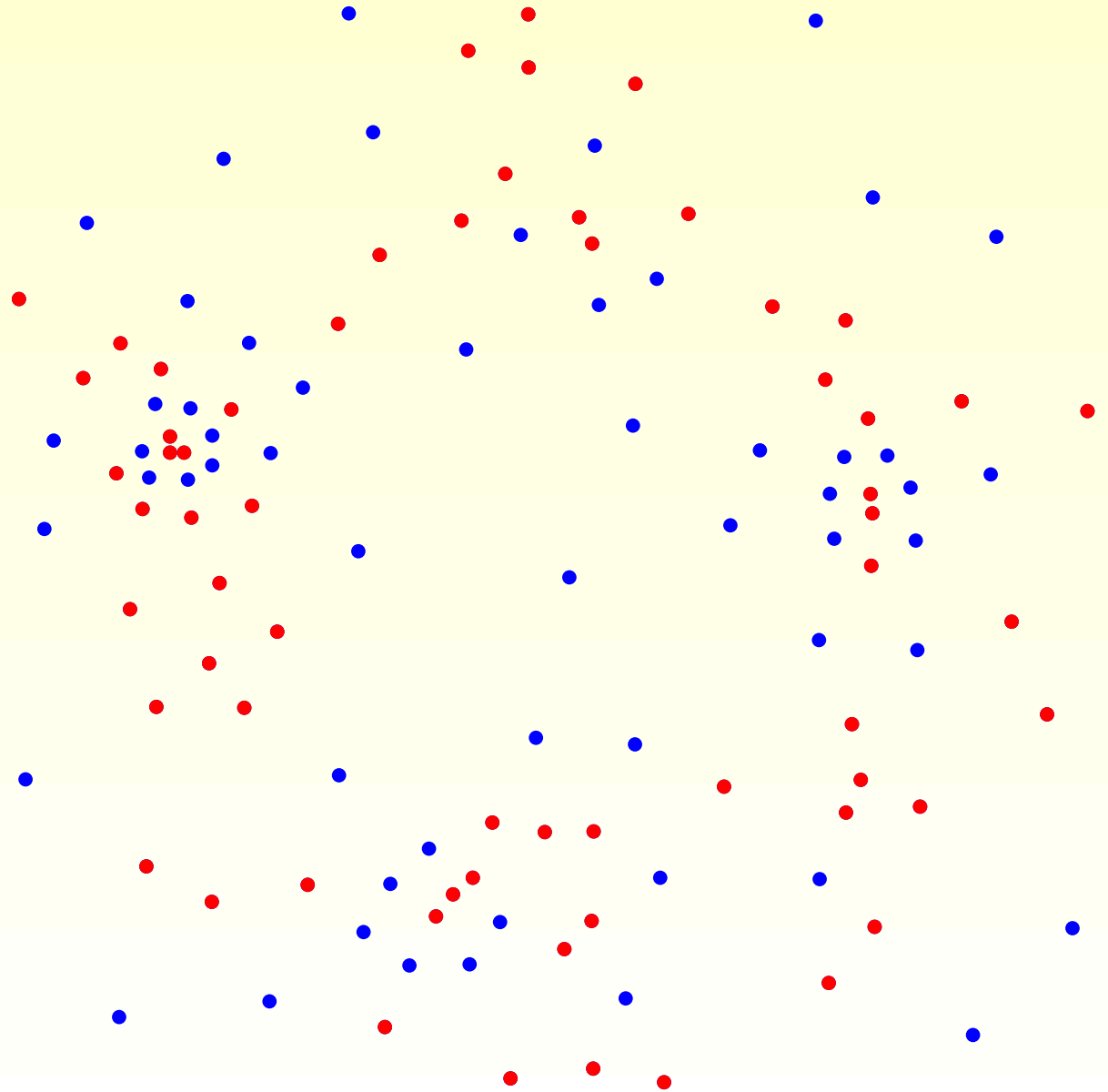
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We need to insert  
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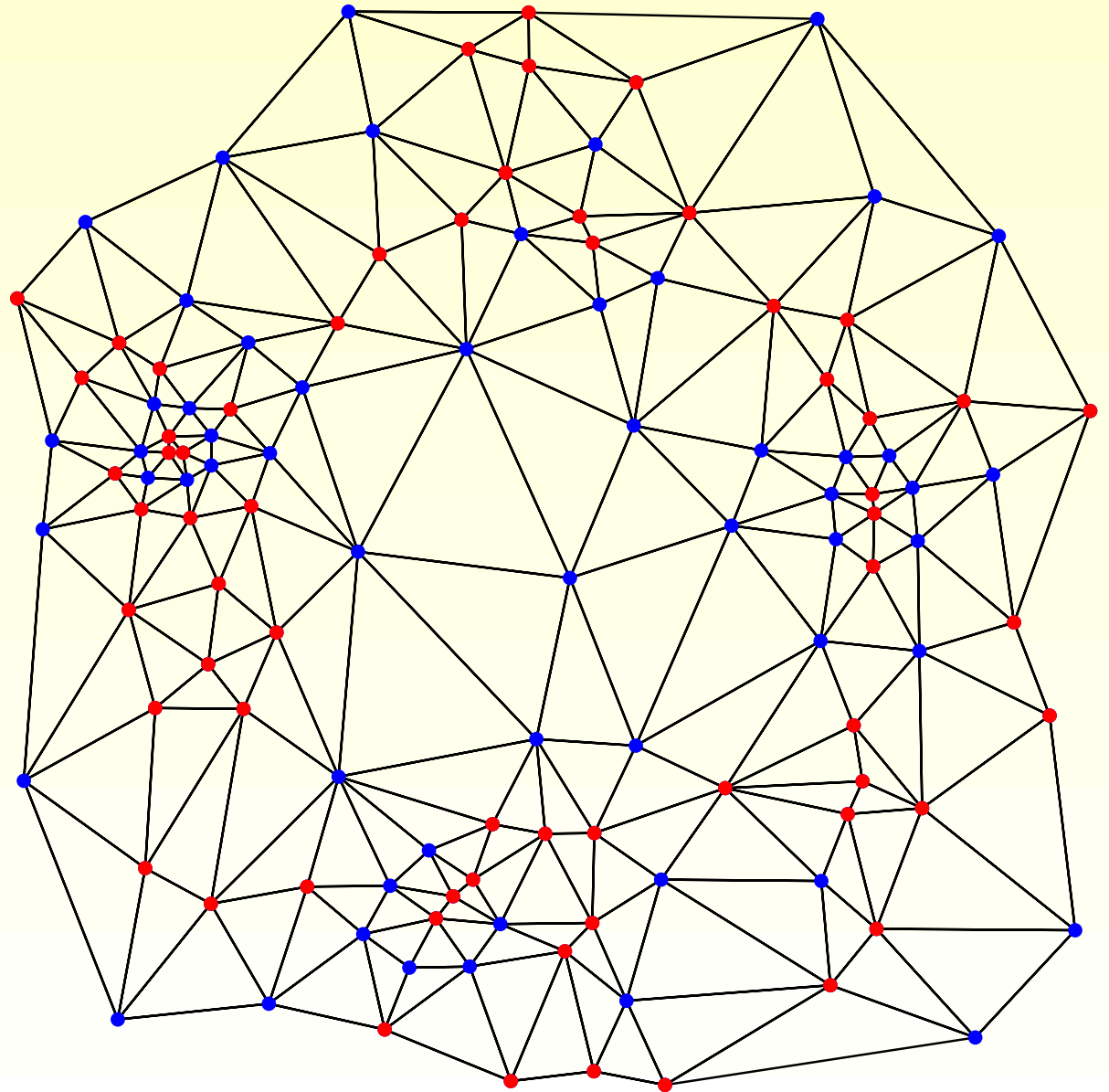


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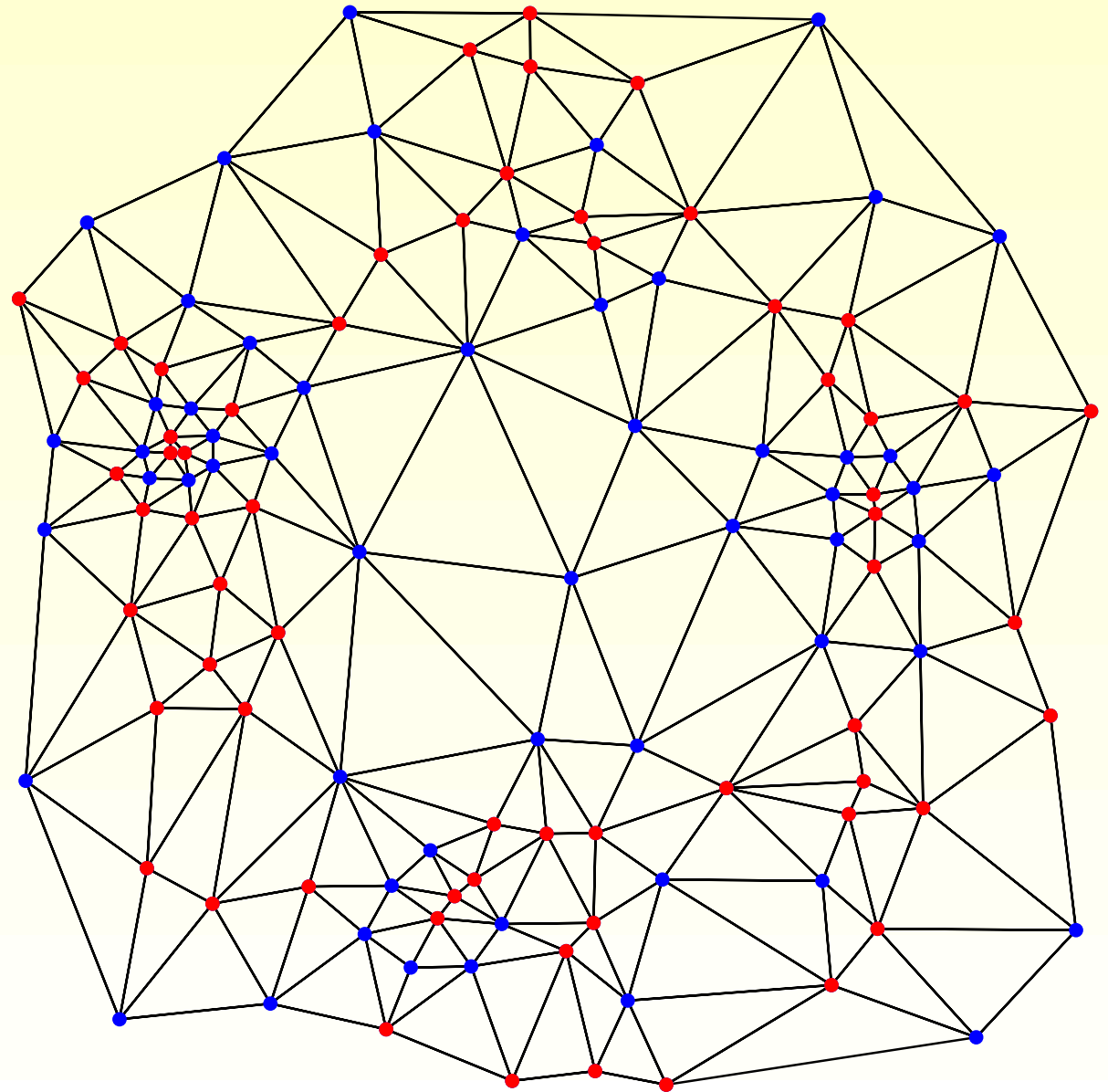
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The output has a  
quality Delaunay  
triangulation, i.e.,  
no small angles



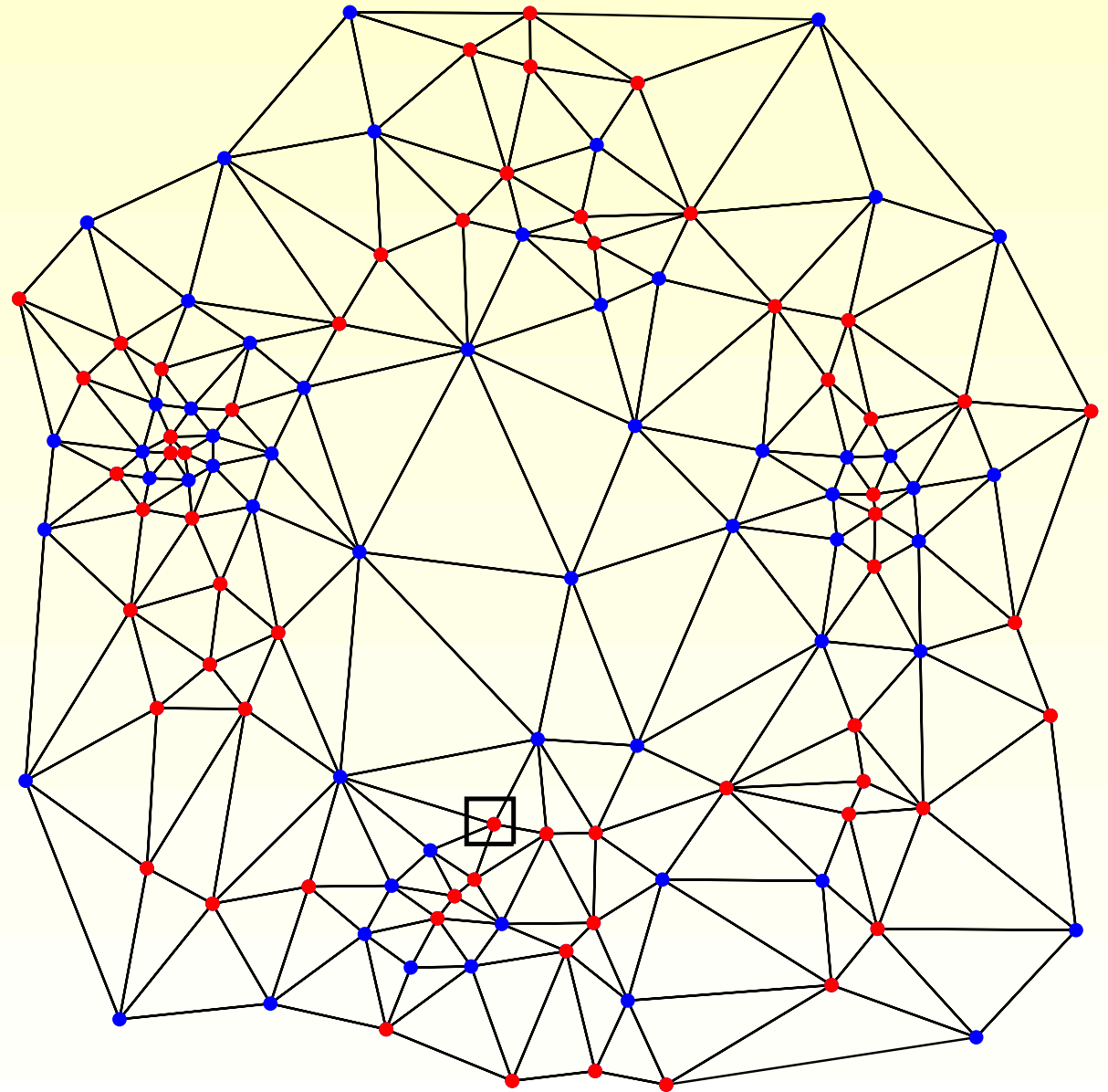
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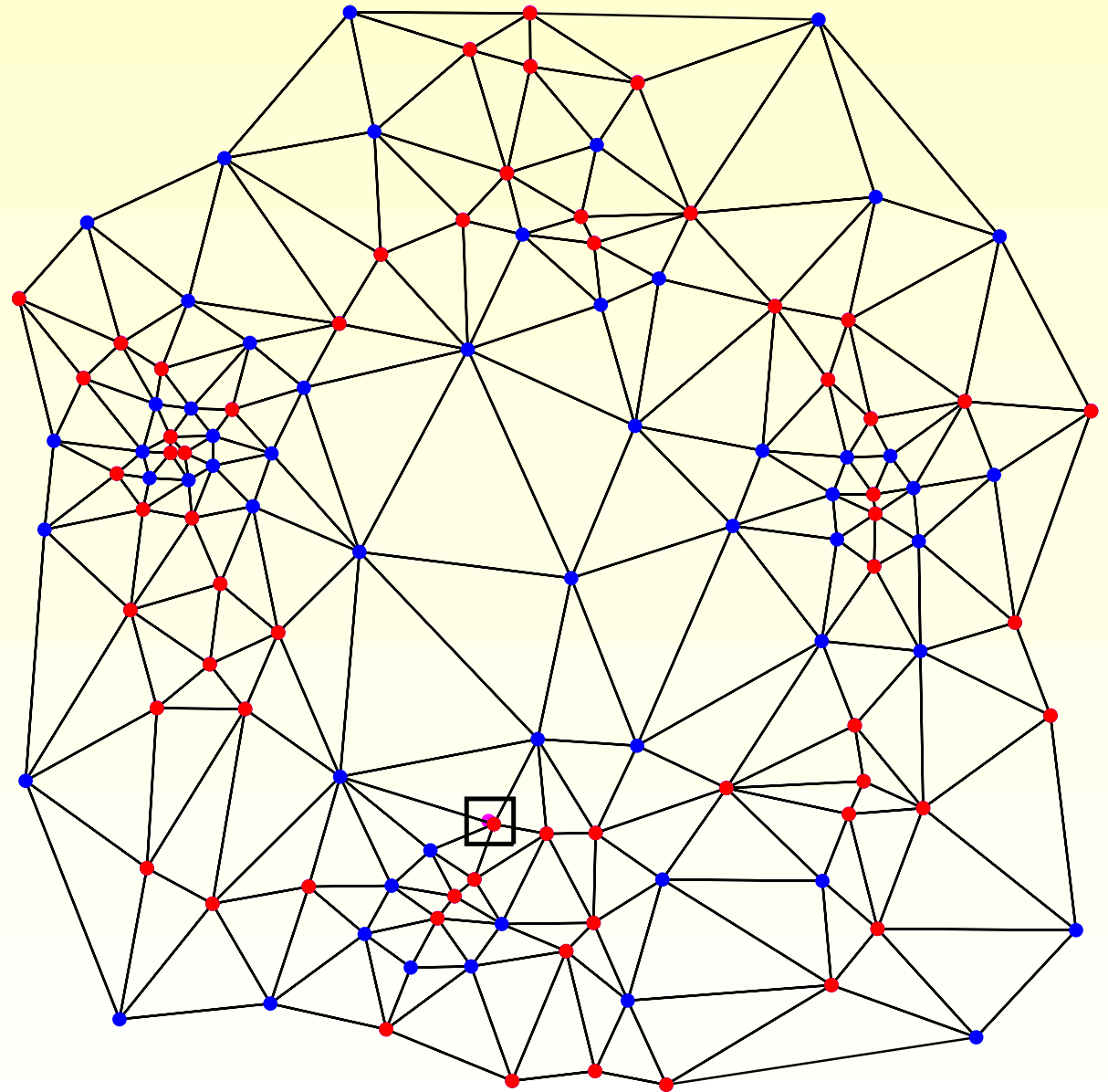




# Dynamic Mesh Refinement

Assume somebody  
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Update the set of  
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efficiently, without  
too many changes

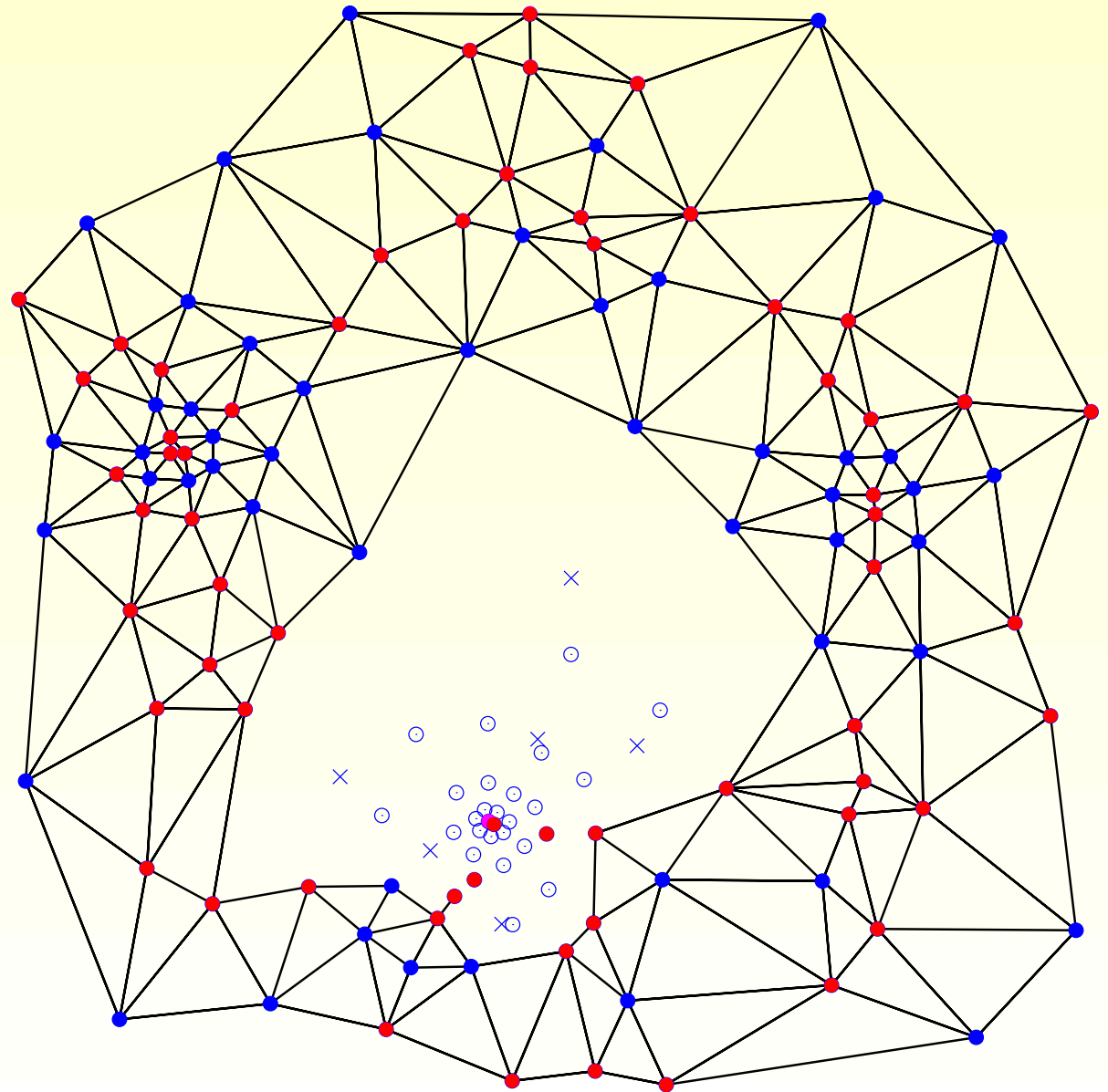


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Assume somebody  
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Sustain output  
size to be small

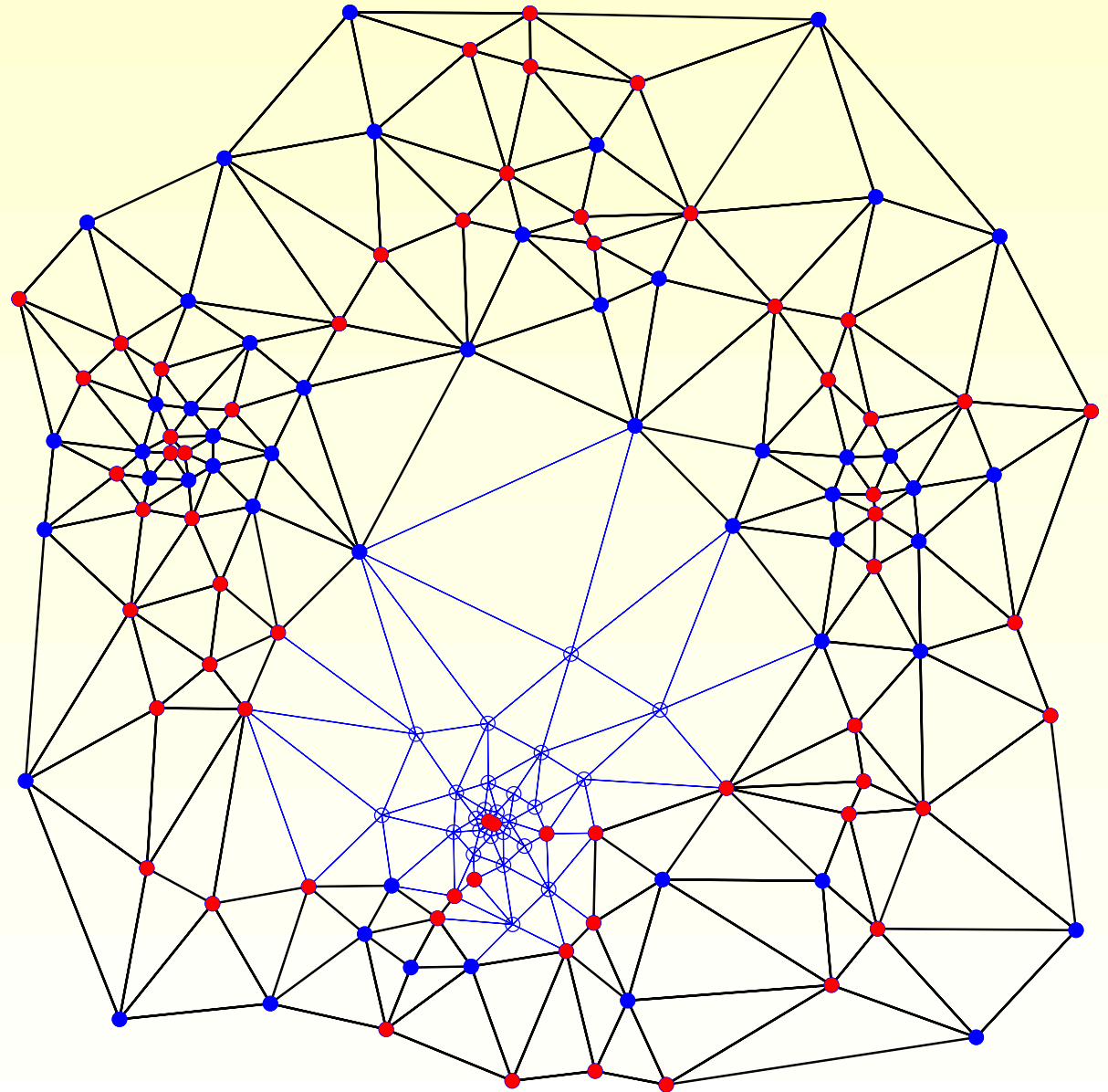


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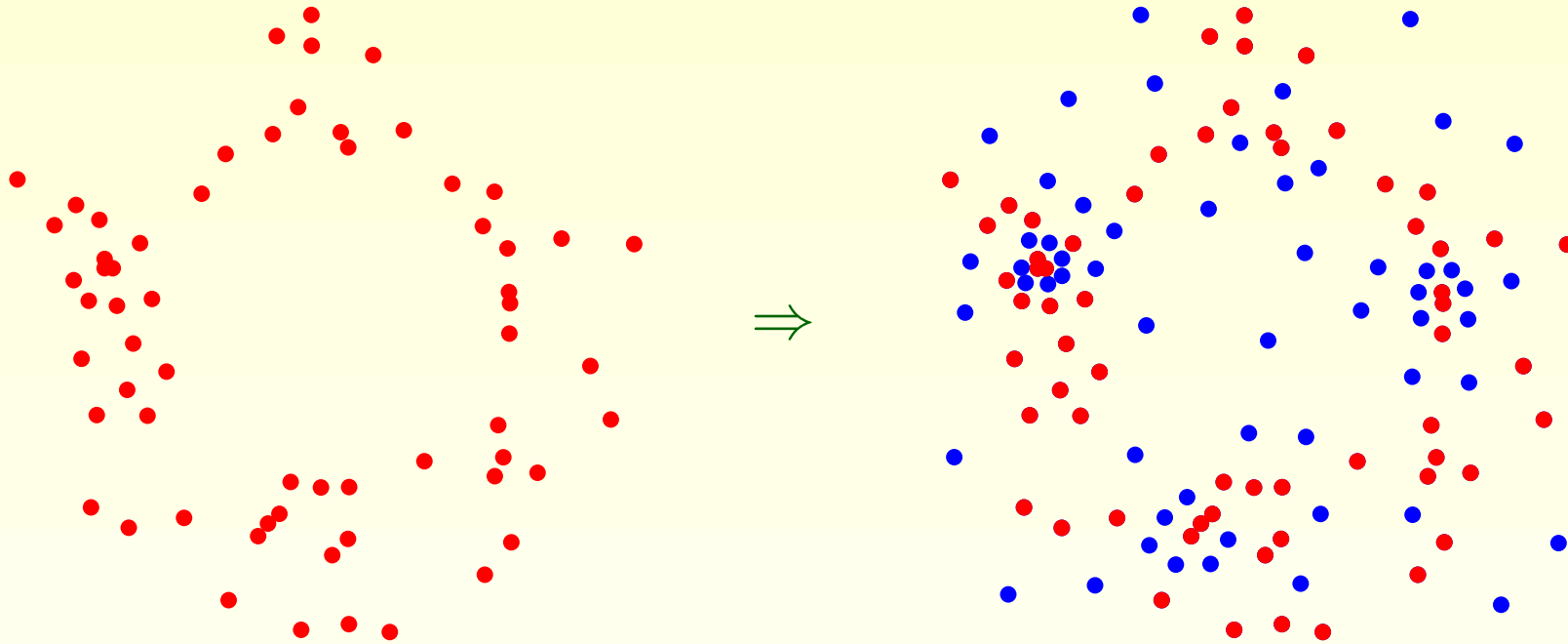
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# Well-Spaced Point Set Problem

Compute a  $\rho$ -well-spaced superset of a given input



Maintain a  $\rho$ -well-spaced superset as input changes

Objectives: **Efficiency** and **Size-Optimality**

# Related Work

- Chew '89 (First non-heuristic construction algorithm)
- Ruppert '95 (Size-optimal algorithm in 2D)
- Spielman, Teng, Üngör '02 (Parallel algorithm in 2D & 3D)
- Har-Peled, Üngör '05 (Efficient algorithm in 2D)
- Hudson, Miller, Phillips '06 (Efficient in arbitrary D)
- Hudson, T '08 (Precursor of this work, arbitrary D)

## Existing Approaches for the Dynamic Problem

- 1) Update is efficient but updated output is not size-optimal
- 2) Worst case update time is as bad as running from scratch

# Our Results

## Construction Algorithm

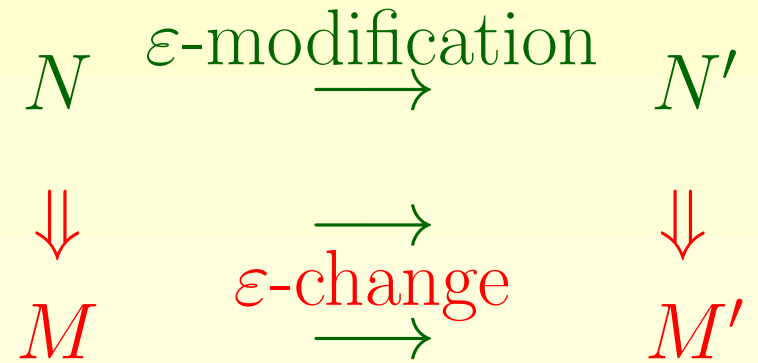
- Given  $N$  and  $\rho > 1$ , output  $M \supset N$  is  $\rho$ -well-spaced
- Output is **size-optimal** w.r.t.  $N$ , i.e.  $|M| = O(|M_{OPT}|)$
- Runs in  $O(n \log \Delta)$  time,  $\Delta = \frac{\text{diameter}}{\text{smallest distance}}$

## Dynamic Update Algorithm

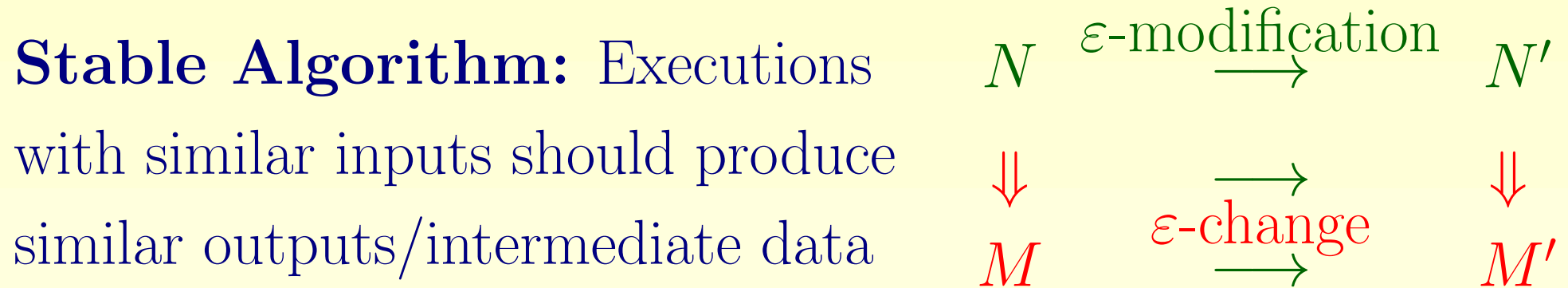
- Given an insertion/deletion ( $N \rightarrow N'$ ), modifies the output
- Modified output  $M'$  is **size-optimal** w.r.t.  $N'$
- Update in  $O(\log \Delta)$  time  $\Rightarrow |M' \ominus M| = O(\log \Delta)$
- Worst case lower bound requires  $|M' \ominus M| = \Omega(\log \Delta)$

# Our Solution — Change Propagation

**Stable Algorithm:** Executions with similar inputs should produce similar outputs/intermediate data



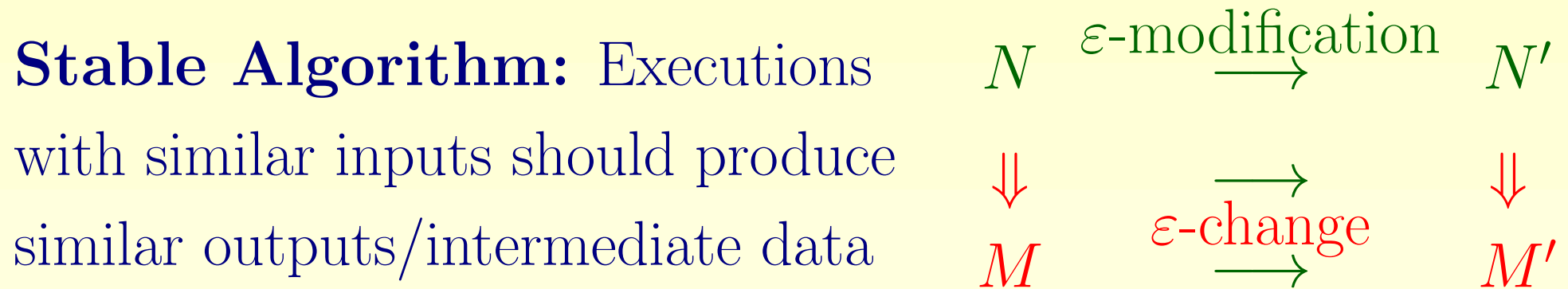
# Our Solution — Change Propagation



**Offline Algorithm  $\Rightarrow$  Quality**



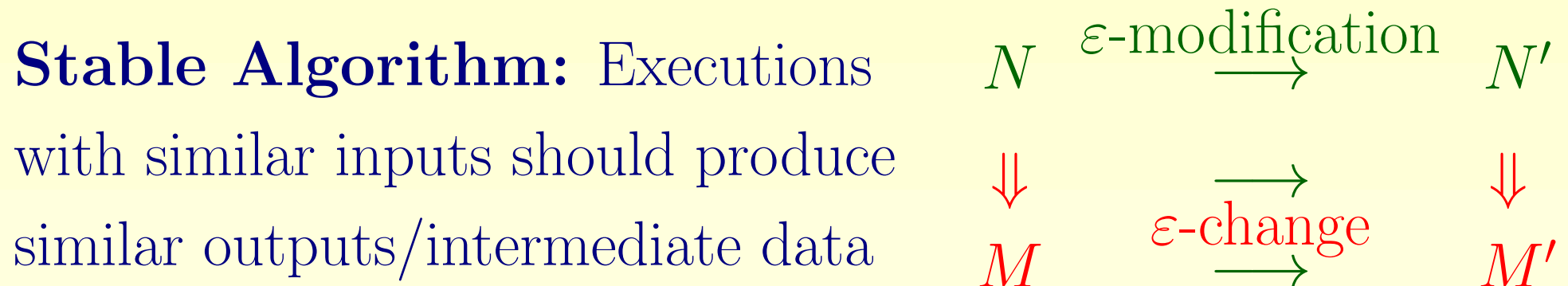
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Offline Algorithm  $\Rightarrow$  Quality

Stability  $\Rightarrow$   $\left\{ \begin{array}{l} \text{Executions are similar} \\ M' \ominus M \text{ is small} \end{array} \right.$

# Our Solution — Change Propagation



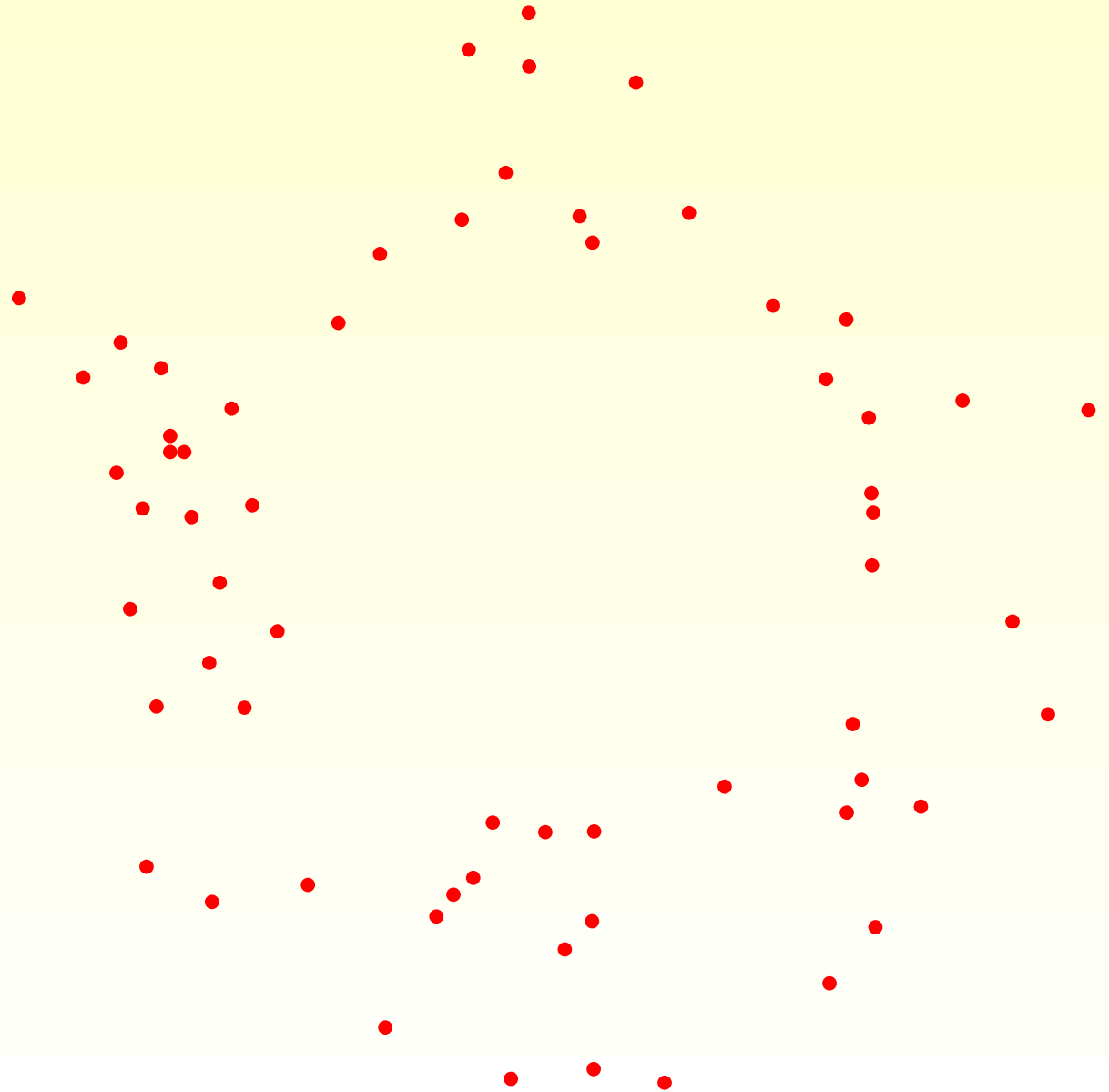
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Dynamizing Stable Algorithm  $\Rightarrow$  Efficiency

# Stable Refinement Algorithm

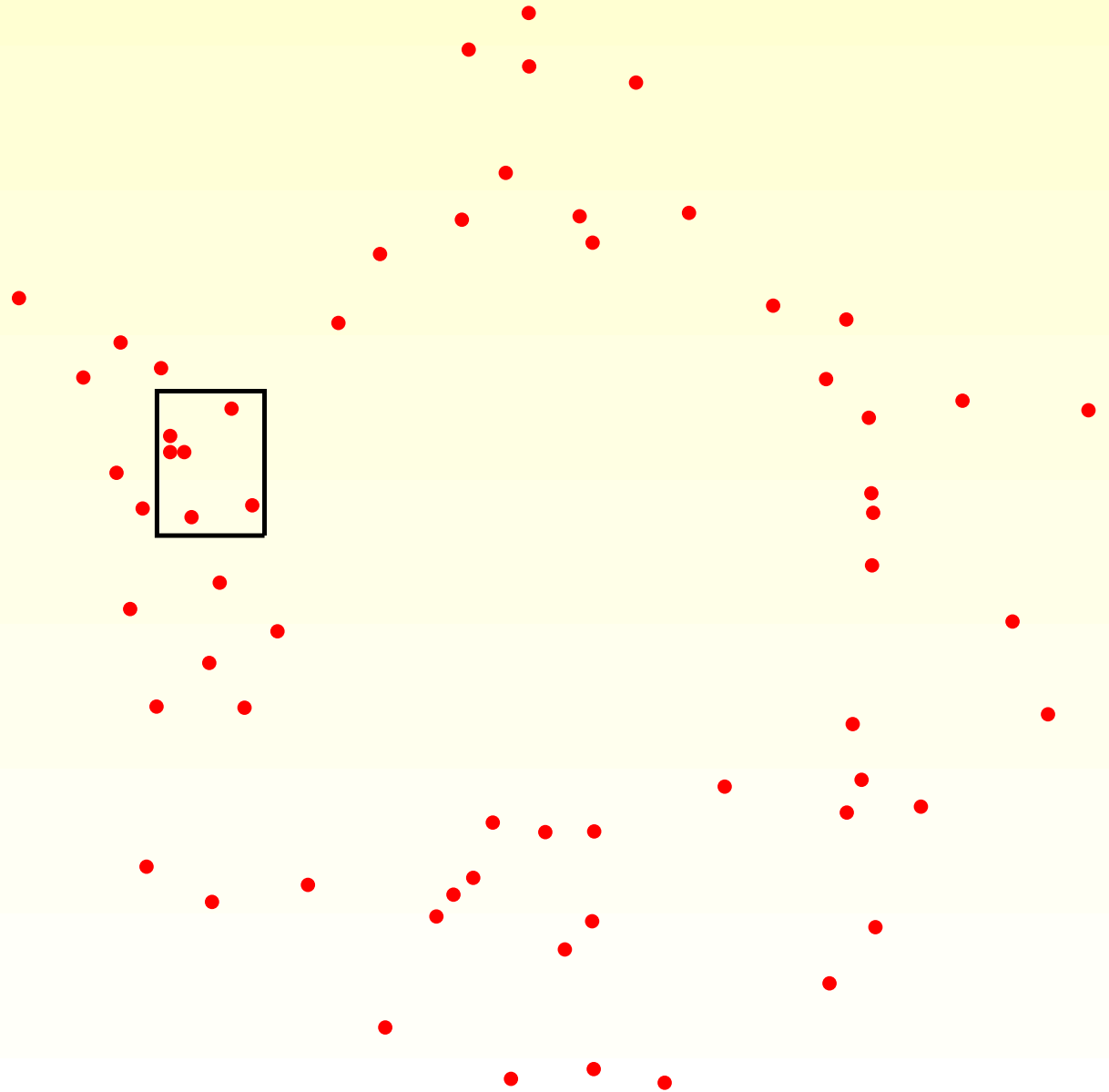
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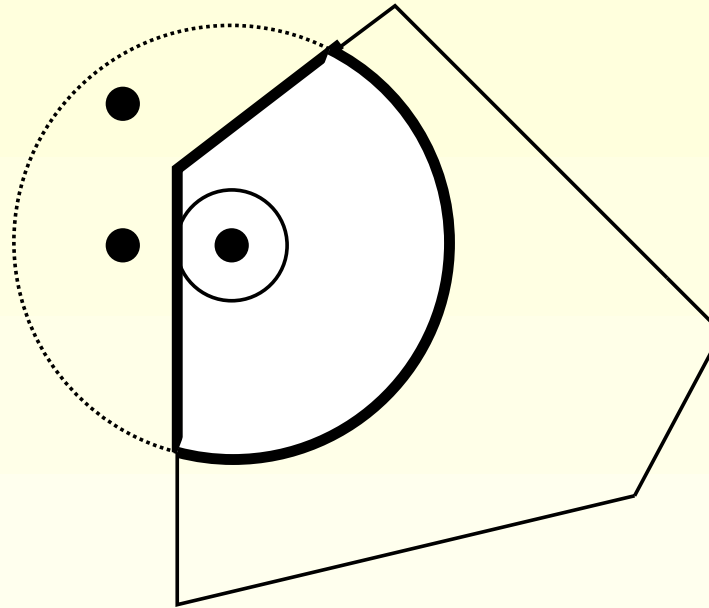
Pick a point that is not well-spaced and fill (insert Steiner points)



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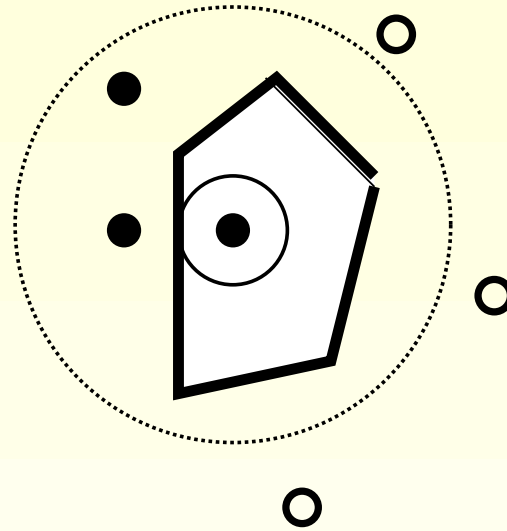
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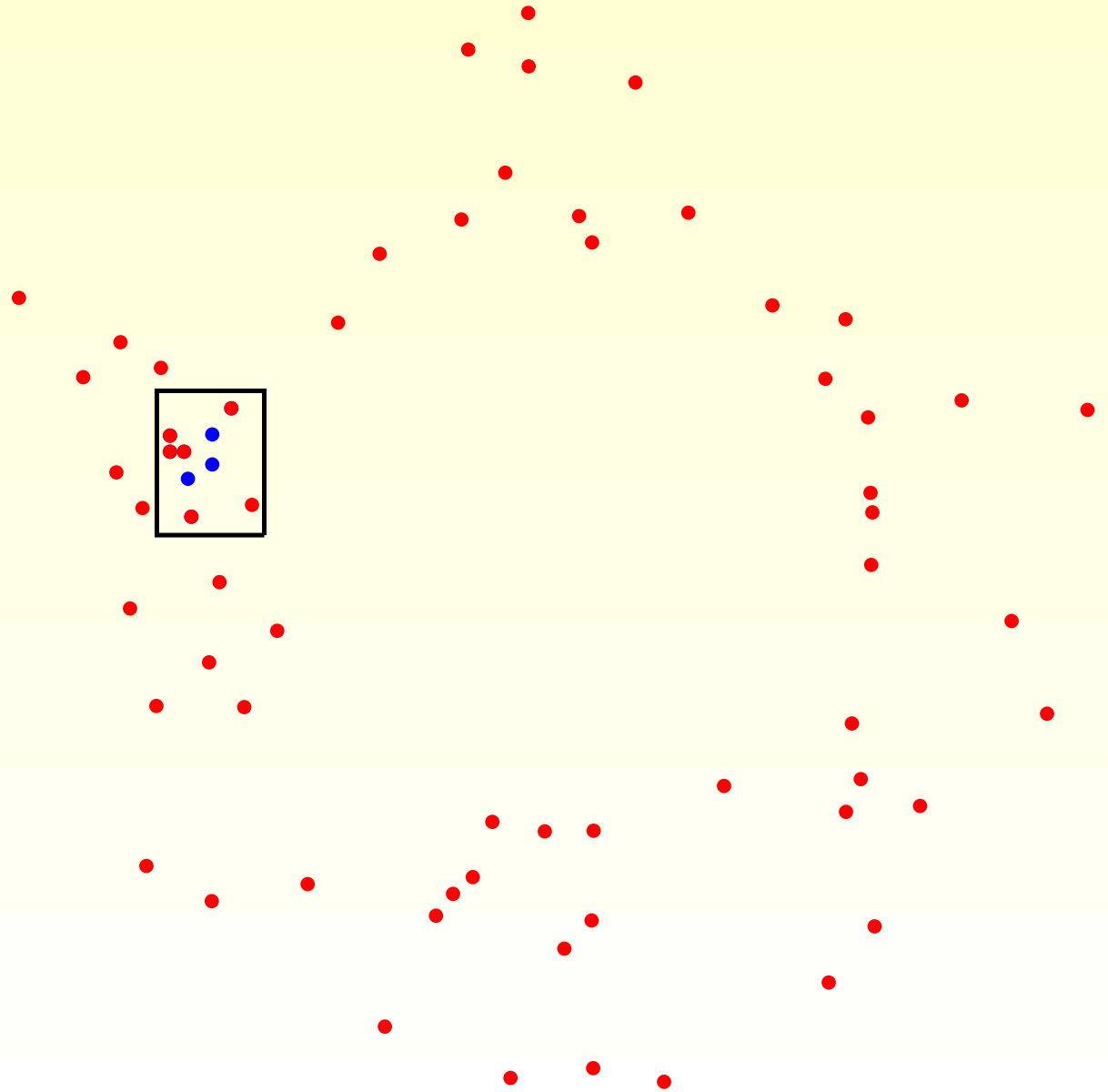


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Idea: Fill points in a **single** pass

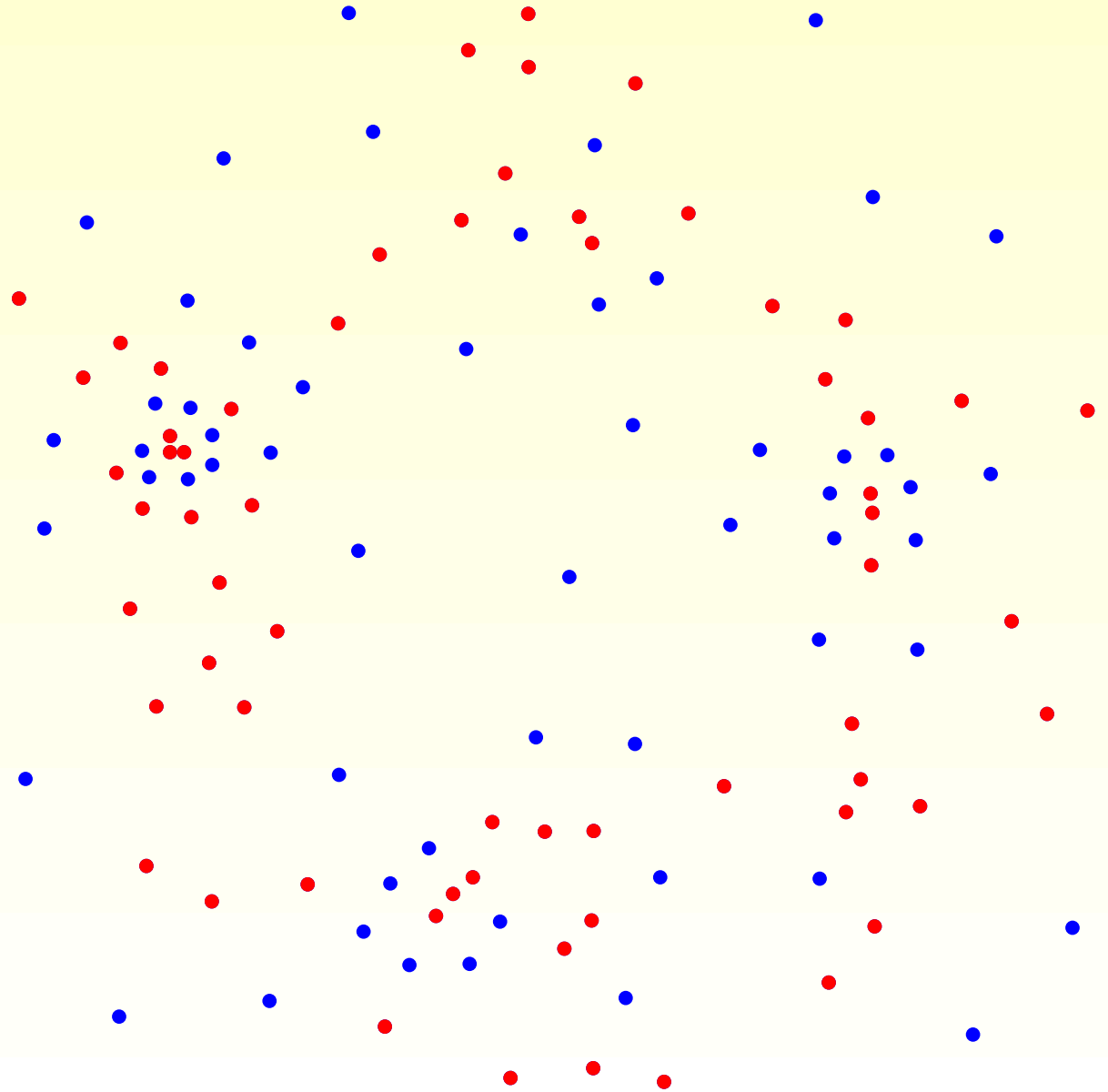


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# Fill Operation and Picking Region

**Goal:** Ensure that filling points in a single pass is sufficient

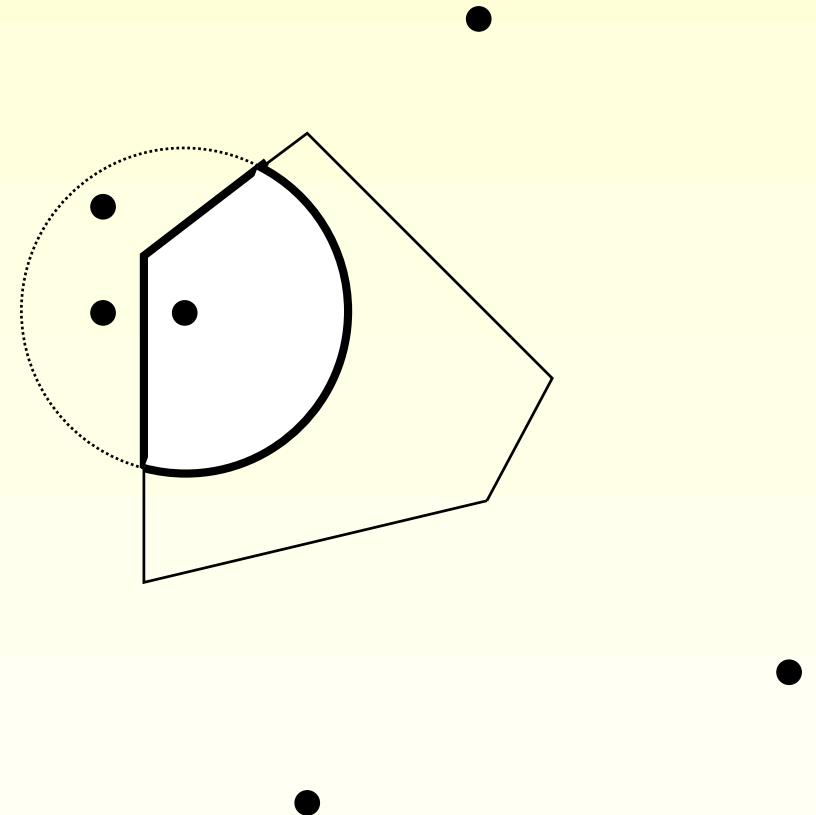
**Approach: Expansion**

Insert Steiner points that have  $\rho$  times larger empty balls

**Goal:** Efficiency, Independence

**Approach: Clipped Voronoi**

Location of Steiner points depend only on close Voronoi neighbors



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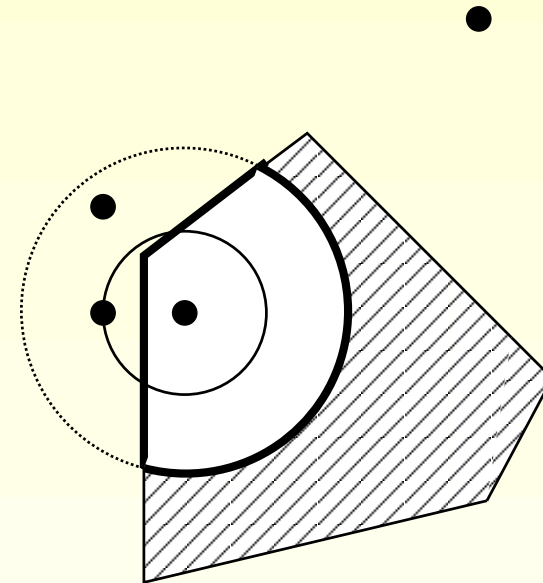
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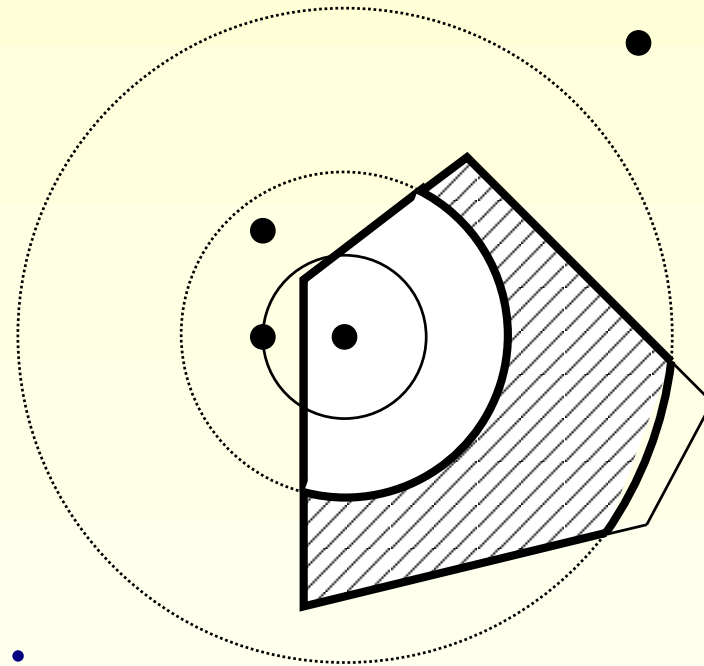
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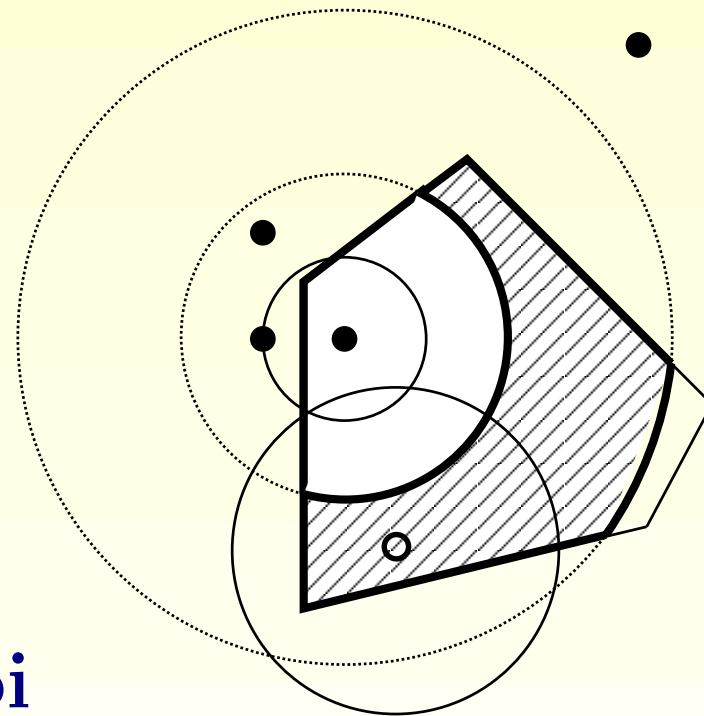
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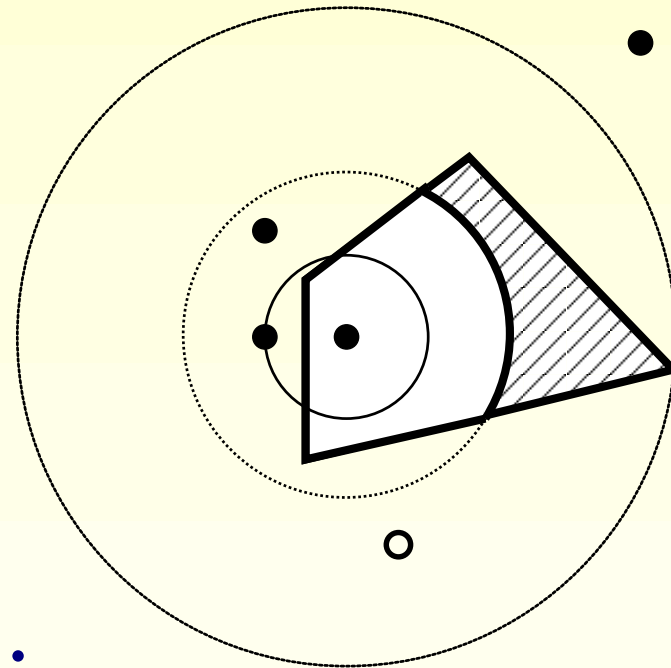
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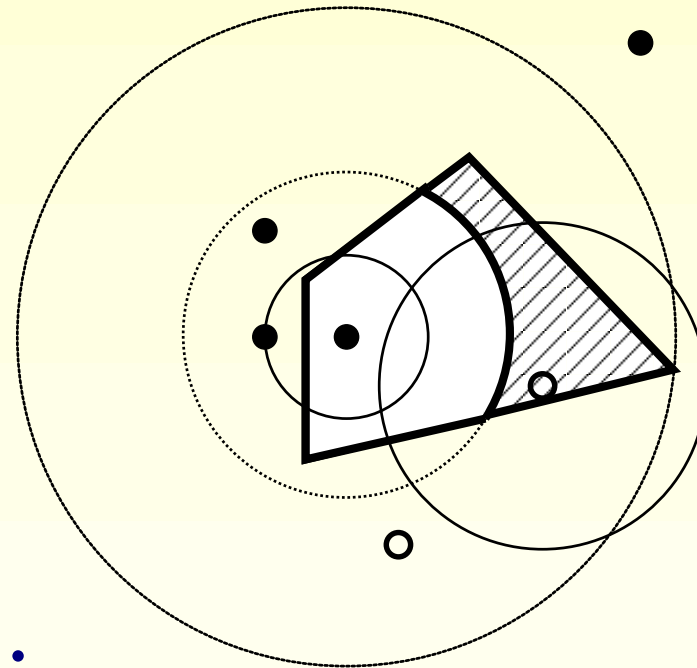
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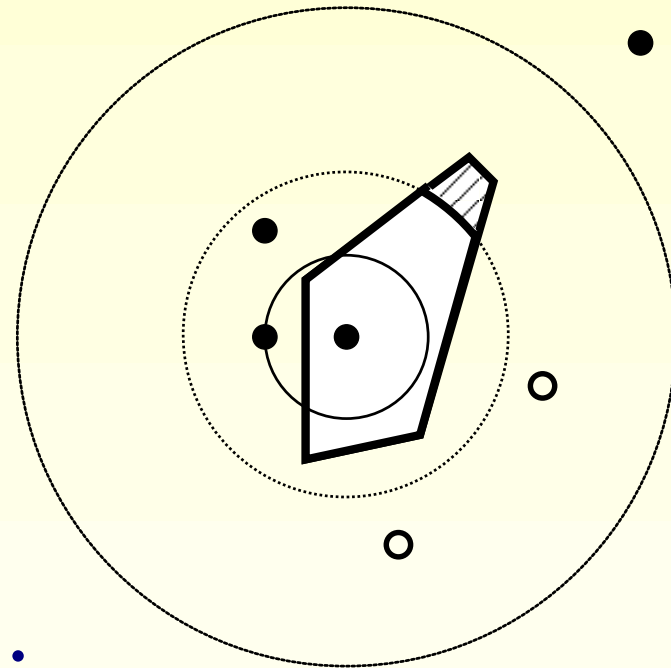
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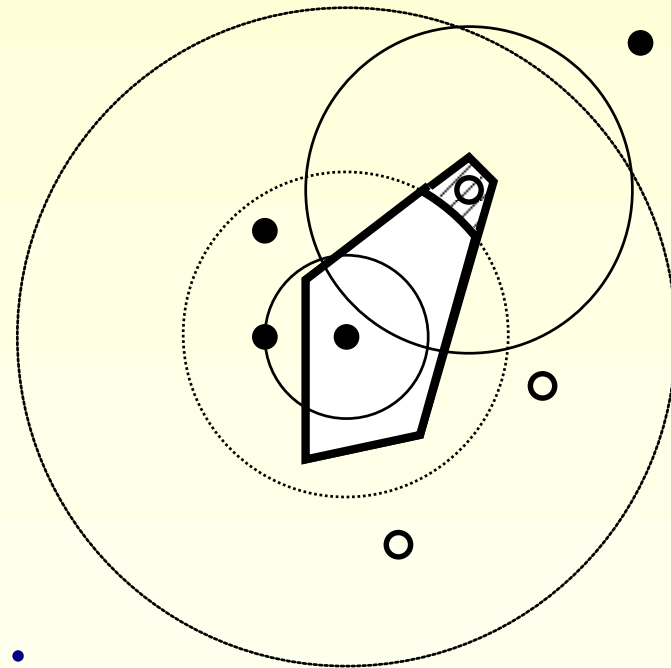
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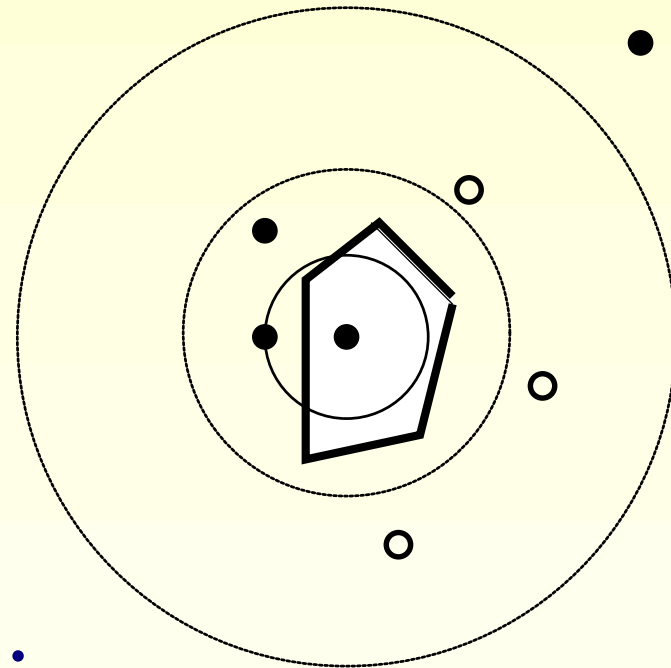
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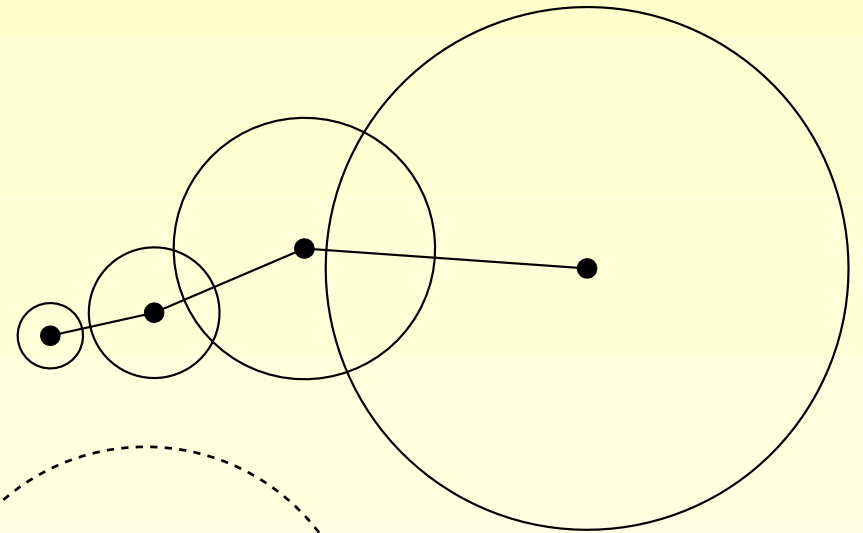
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# Key Ideas for Stability

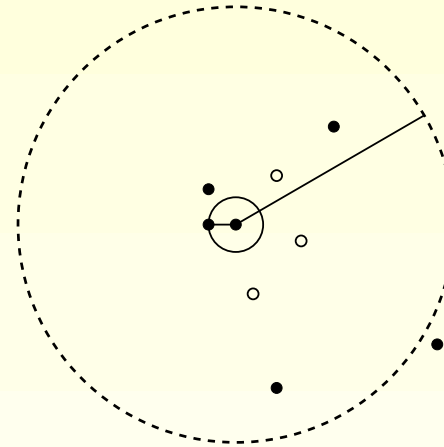
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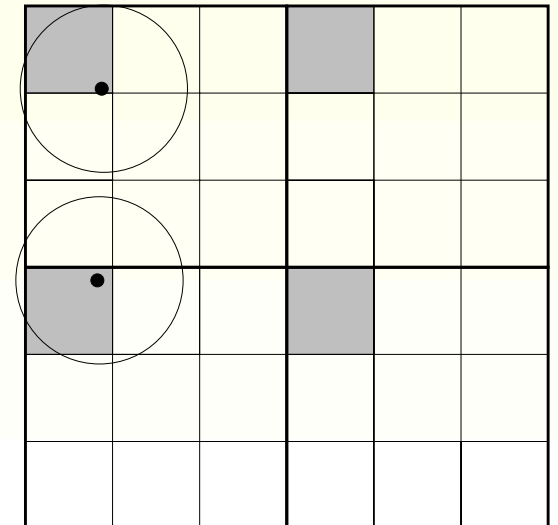
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Steiner point selection depends on a bounded neighborhood



## Independence

At each rank, filling distant points are guaranteed not to affect each other



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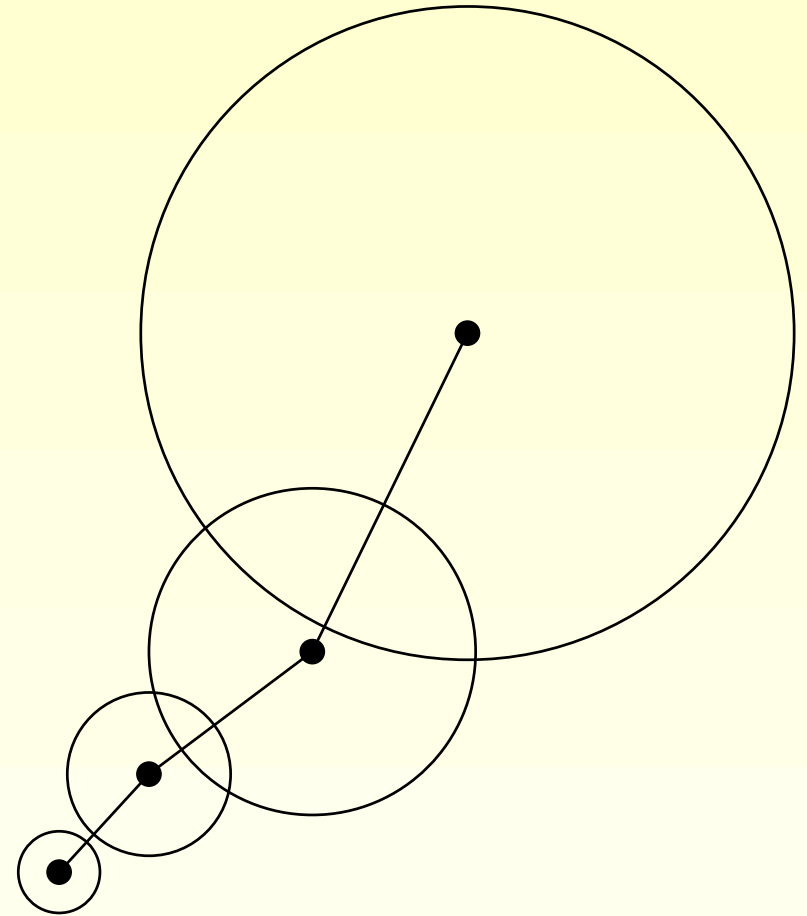
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**Rank** = logarithm base  $\rho$   
of nearest neighbor distance  
In total  $O(\log \Delta)$  ranks

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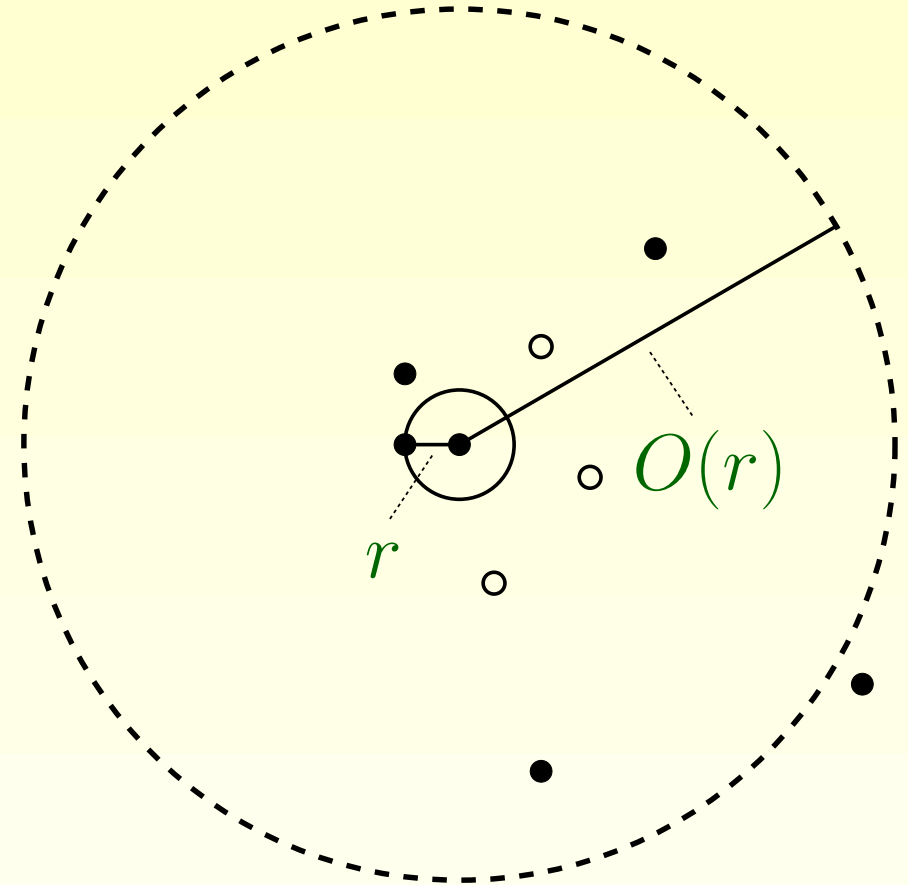
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Steiner point selection depends on a bounded neighborhood

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Points outside the big ball do not influence how we pick Steiner points

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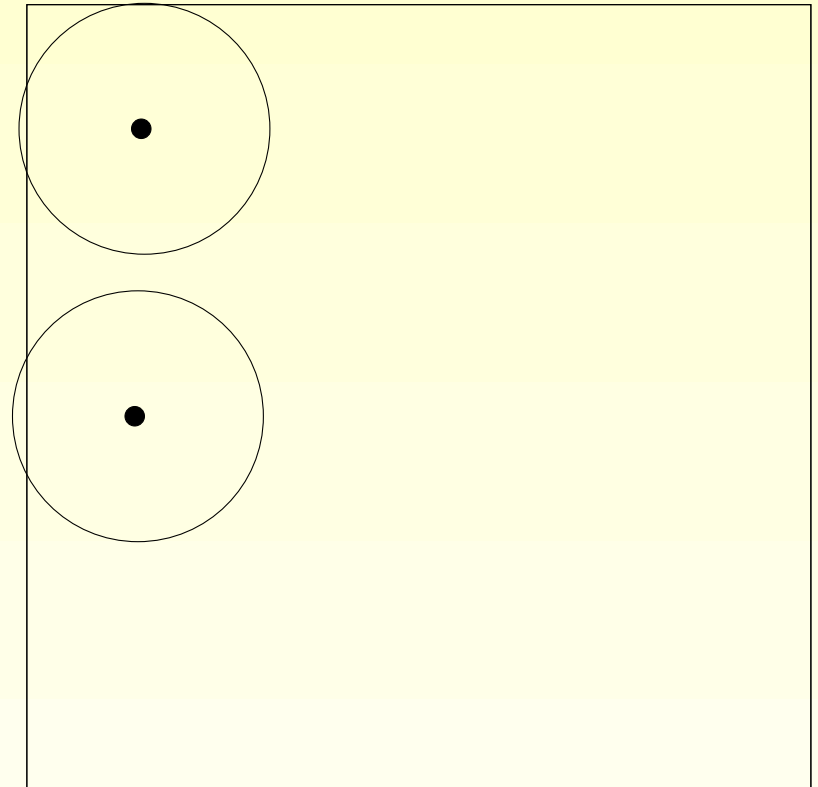
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Process points in rank order

Two points at a given rank may not depend on each other

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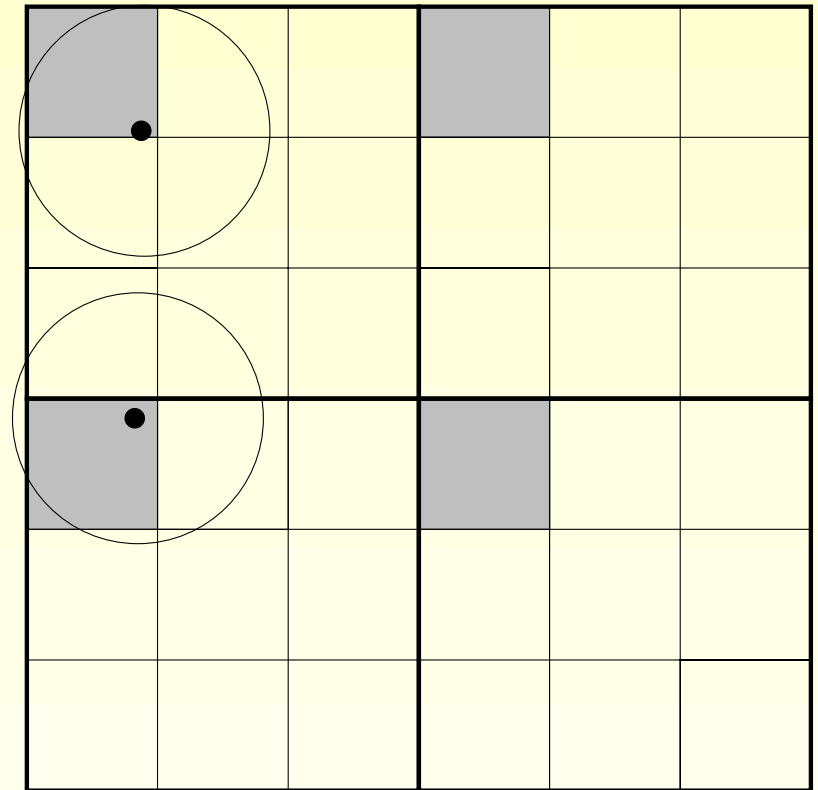
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For ensuring independence partition space using  $O(1)$  colors and process points in color order

# Proof of Stability — Dependency Path

## Schedule using Ranks and Colors

Algorithm processes points in **stages** by ordering them first by rank then by color

## Dependency Paths

Expansion and independence guarantee that filling a point does not affect points processed in earlier stages and the points being processed at the current stage

**There are  $O(\log \Delta)$  stages in total**

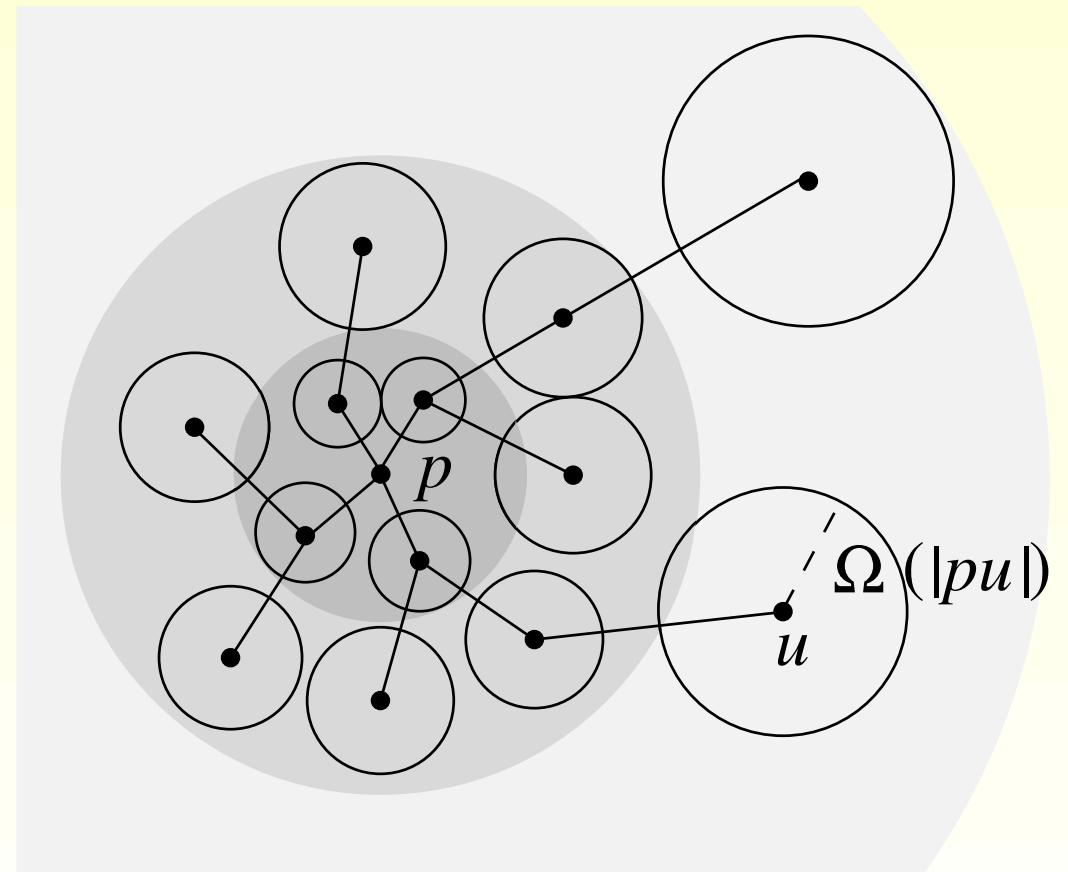
# Proof of Stability — Spacing and Packing

Assume inserting (or deleting) a point  $p$

**Path:** Given a point  $u$  at rank  $r$  if there is a dependency path from  $p$  to  $u$ , then  $|pu| < O(\rho^r)$

**Spacing:** There is an empty ball around  $u$  of radius  $\Omega(\rho^r)$  which consequently is  $\Omega(|pu|)$

**Packing:** There can be only  $O(1)$  such points at each rank  $O(\log \Delta)$  in total





Thank You!

Questions?