## Closest Points

- Closest Pair
- Nearest Neighbor



## Closest Pair

## Given a set P of N points, find $\mathrm{p}, \mathrm{q} \in \mathrm{P}$ such that the distance $d(p, q)$ is minimum.


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- Algorithms for determining the closest pair:
- brute force $\mathrm{O}\left(\mathrm{N}^{2}\right)$
- divide-and-conquer $\mathrm{O}(\mathrm{N} \log \mathrm{N})$
- plane-sweep $O(N \log N)$


## Brute Force Algorithm

## Compute all the distances $\mathrm{d}(\mathrm{p}, \mathrm{q})$ and select the minimum distance.

$$
\begin{aligned}
& \mathrm{p}_{1}\left(\mathrm{x}_{1}, \mathrm{y}_{1}\right) \\
& \mathrm{d}\left(\mathrm{p}_{1}, \mathrm{p}_{2}\right)=\sqrt{\left(\mathrm{x}_{2}-\mathrm{x}_{1}\right)^{2}+\left(\mathrm{y}_{2}-\mathrm{y}_{1}\right)^{2}} \\
& \text { Time Complexity: } \mathbf{O}\left(\mathbf{N}^{2}, \mathrm{y}_{2}\right)
\end{aligned}
$$

## Plane-Sweep Algorithm

- Maybe we can avoid having to check the distance between every pair of points...
- Plane-sweep worked for segment intersection, maybe it can be useful here...
- Key observation: if the closest pair of points to the left of the sweep line is distance $d$ apart, the next point encountered can't be a closest pair with any point more than $d$ units to the left of the line



## Stored Information

- Maintain the following information:
- the closest pair $(a, b)$ found so far, and the distance d between them
- ordered dictionary $S$ of the points lying in a strip of width $d$ to the left of the sweep line, using the $y$ coordinates as keys



## Updating

- When the sweep line encounters a point p:
- update the dictionary so it only contains points that might be a closest pair with p
- remove all points $r$ such that $x(p)-x(r)>d$ from $S$
- find the closest point $q$ to $p$ in $S$
- if $d(p, q)<d$ then update the current closest pair and distance
- insert p into $S$



## Searching the Dictionary

- How to quickly find the closest point in the dictionary?
- could be $\mathrm{O}(\mathrm{N})$ points in the dictionary...
have $\mathrm{x}, \mathrm{y}$ spacing so that $\mathrm{y}=\mathrm{d} /(\mathrm{n}-1)$
- Good news: not all of the points in the dictionary can improve d
- only eligible points are in half circle of radius $d$ centered at $p$



## Searching the Dictionary II

- But how to search in a half-circle?
- a rectangle is almost a half-circle...
- do a range search in the interval [y(p)-d,y(p)+d]
- this will get all the points in the half-circle (and maybe some others)

- Use brute-force to check the distance to each point returned by the range query
- But isn't that still a potentially large number of points?
- actually, there are at most 6
- key observation: all of the points in the dictionary are at least distance d from each other



## Putting It All Together

- sort points by x-coordinate and store in ordered sequence $X$
- maintain references to two positions in sequence
- firstInStrip: the leftmost point in $S$
- lastInStrip: the new point to be added to $S$
- at each step..
// advance lastInStrip
lastlnStrip $\leftarrow$ X.after(lastlnStrip)
// remove points that are no longer candidates from dictionary
while $\times($ point(firstlnStrip)) $<x$ (point(lastlnStrip))-d do
S.remove(point(firstlnStrip))
firstInStrip $\leftarrow$ X.after(firstInStrip)
// update closest point information
find point q closest to point(lastlnStrip) in S
if $d(p, q)<d$ then
update closest pair
$\mathrm{d} \leftarrow \mathrm{d}(\mathrm{p}, \mathrm{q})$
// insert new point into dictionary
S.insert(point(lastInStrip))


## An Example




one point in rectangle but not half-circle; closest pair not updated

## An Example Continued


one point in rectangle but not half-circle; closest pair not updated
$\bullet$

one point in rectangle but not half-circle; closest pair not updated

## Still Going...


two points in rectangle, one on border of half-circle; closest pair not updated

two points in rectangle and half-circle; closest pair updated to nearer of the two

## Example Completed


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$$
\begin{aligned}
& \text { • } \\
& \\
& \\
& \\
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& \\
& \text { ene final result, with closest } \\
& \text { pair shown }
\end{aligned}
$$

## Running Time

- initial sort takes $\mathrm{O}(\mathrm{N} \log \mathrm{N})$ time
- each point is inserted and removed once from $S$
- S has at most N elements, so each insertion/ removal takes $\mathrm{O}(\log \mathrm{N})$ time
- total insertion/removal time is $\mathrm{O}(\mathrm{N} \log \mathrm{N})$
- dictionary is searched once each time a point is inserted into $S$
- each range query takes $\mathrm{O}(\log \mathrm{N}+6)=\mathrm{O}(\log \mathrm{N})$ time
- total time for range queries is $\mathrm{O}(\mathrm{N} \log \mathrm{N})$
- distance computations performed each time a point is inserted into S
- at most 6 computations at each time
- total time for distance computations is $\mathrm{O}(\mathrm{N})$


## Time Complexity: $\mathbf{O}(\mathbf{N} \log N)$ (definitely beats the brute force method!)

## Nearest Neighbor

- Given a set $S$ of sites, what is the closest site to point q?

- Brute force is only $\mathrm{O}(\mathrm{N})$ !
- but if you repeat the query for $k$ different points (using the same set of sites) the total time is $\mathrm{O}(\mathrm{kN})$
- Could do something based on plane-sweep, but that takes $\mathrm{O}(\mathrm{N} \log \mathrm{N})$ time for each query...O $(\mathrm{kN} \log \mathrm{N})$ for k queries
- There's a better solution...


## Voronoi Diagram

- $S=\left\{s_{1}, s_{2}, \ldots, s_{N}\right\}$
- set of points in the plane, called sites
- Voronoi cell of $\mathrm{s}_{\mathrm{i}}$ :
$-\mathrm{C}\left(\mathrm{s}_{\mathrm{i}}\right)=\left\{\mathrm{p}: \mathrm{d}\left(\mathrm{p}, \mathrm{s}_{\mathrm{i}}\right) \leq \mathrm{d}\left(\mathrm{p}, \mathrm{s}_{\mathrm{j}}\right), \forall \mathrm{j} \neq \mathrm{i}\right\}$
- that is, the region of the plane containing all of the points that are closer to $\mathrm{s}_{\mathrm{i}}$ than any other site $\mathrm{s}_{\mathrm{j}}$
- Voronoi diagram of S
- subdivison of the plane into Voronoi cells



## Constructing a Voronoi Diagram



- Construct the perpendicular bisectors $\mathrm{h}_{\mathrm{ij}}$ of each segment $\left(\mathrm{s}_{\mathrm{i}}, \mathrm{s}_{\mathrm{j}}\right)$
- Let $\mathrm{H}_{\mathrm{ij}}$ be the half-plane delimited by $\mathrm{h}_{\mathrm{ij}}$ and containing $\mathrm{s}_{\mathrm{i}}$
- all the points p in $\mathrm{H}_{\mathrm{ij}}$ are closer to $\mathrm{s}_{\mathrm{i}}$ than $\mathrm{s}_{\mathrm{j}}$
- Voronoi cell for $\mathrm{s}_{\mathrm{i}}$ is the intersection of the halfplanes $\mathrm{H}_{\mathrm{ij}}$ for all sites $\mathrm{s}_{\mathrm{j}}(\mathrm{j} \neq \mathrm{i})$
- Voronoi diagram can be constructed in $\mathrm{O}(\mathrm{N} \log \mathrm{N})$ time
- can use divide-and-conquer or plane-sweep technique


## Fun Voronoi Facts

- Each Voronoi cell is convex
- A Voronoi cell is unbounded if and only if the site is on the convex hull
- If $\mathrm{s}_{\mathrm{j}}$ is the nearest neighbor of $\mathrm{s}_{\mathrm{i}}$, the Voronoi cells $\mathrm{C}\left(\mathrm{s}_{\mathrm{i}}\right)$ and $\mathrm{C}\left(\mathrm{s}_{\mathrm{j}}\right)$ touch



## Applications

- Given the Voronoi diagram, a nearest neighbor query can be performed in $\mathrm{O}(\log \mathrm{N})$ time
- k queries can be done in $\mathrm{O}((\mathrm{N}+\mathrm{k}) \log \mathrm{N})$ time
- Other applications
- all nearest neighbors: for every point $p \in P$, find its nearest neighbor $q$
- closest pair
- Delaunay triangulation
- a triangulation is a division of the plane into a set of triangular regions
- convex hull
- not just limited to computational geometry...
- model region of influence in archaeology, ecology, ...


## Shameless Plug

- Want to know how to compute a Voronoi diagram in $\mathrm{O}(\mathrm{N} \log \mathrm{N})$ time?
- Want to know how to do a nearest-neighbor query in $\mathrm{O}(\log \mathrm{N})$ time?
- Want to learn about other cool geometric algorithms?


# Take CS252: <br> Computational Geometry! 

