

Insertion Sort  
and  
Closest Features  
A Novel Approach to Collision  
Detection

Alexander Schinner  
Otto-von-Guericke Universitt Magdeburg,  
Germany

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The basic idea:  
Information is expensive!



MD-Simulation of granular materials



**slowly** changing system



The last time-step's information is **nearly** correct.



Reuse this information as **good** starting conditions.

## The steps of a MD-simulation

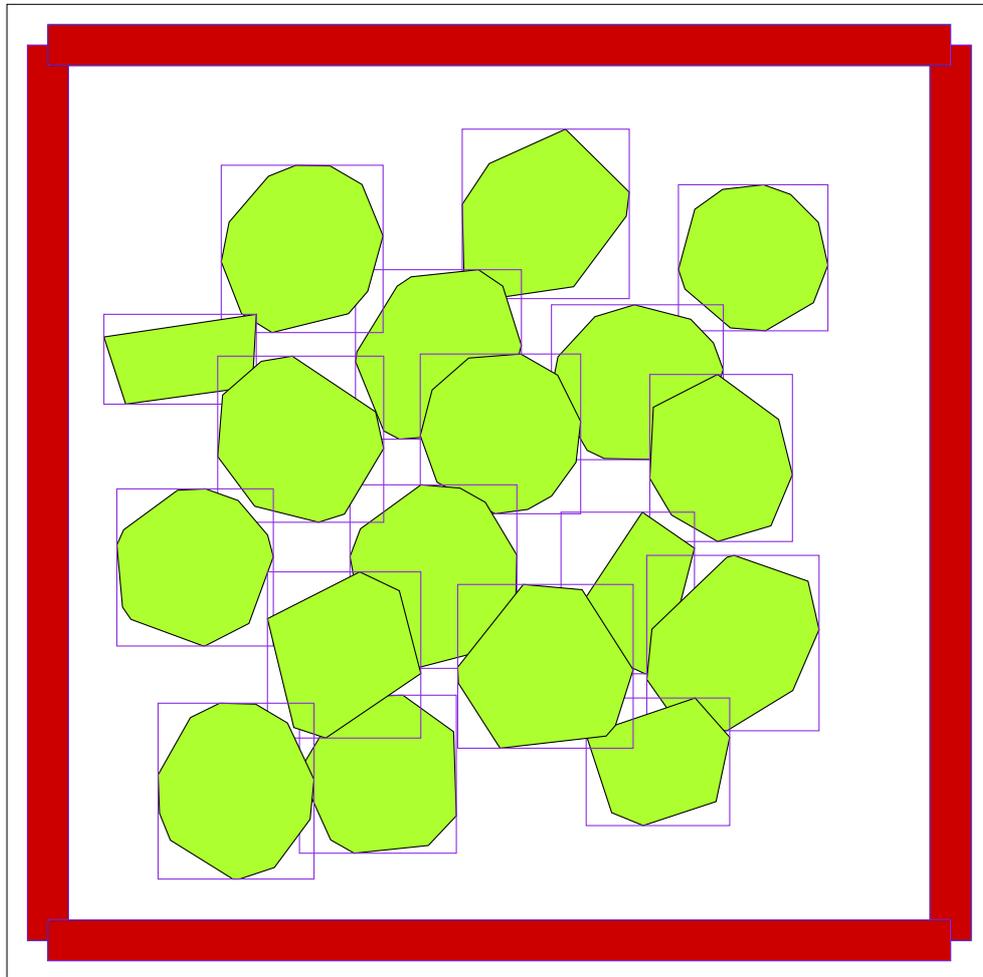
1. Check for bounding box collisions
2. Check for particle collisions
3. Calculate overlap
4. Do the physics (not discussed here)

Increasing  
need of  
time



Try to **exclude** as many particle-pairs as possible from being considered in the next level.

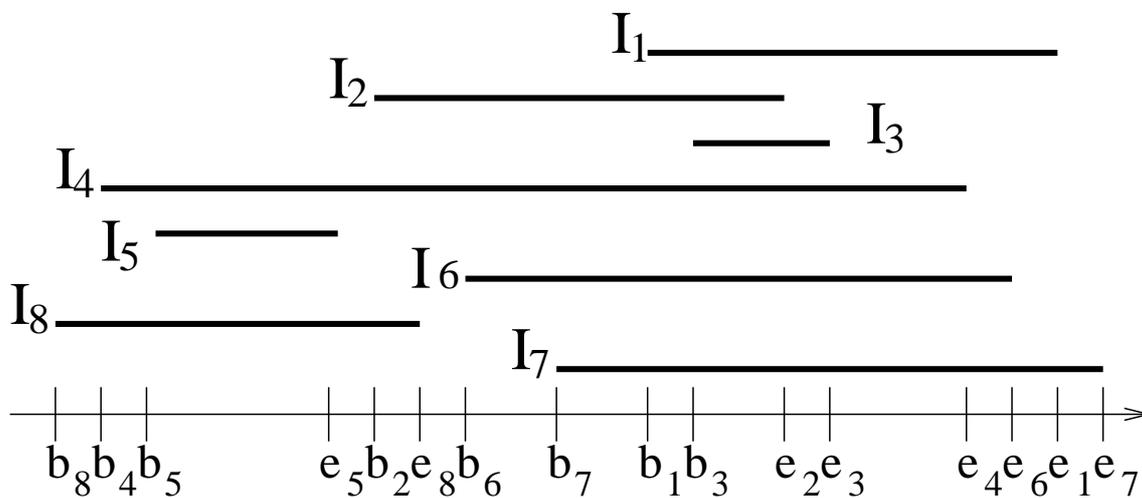
## Bounding Boxes



Pairwise check  $\Rightarrow N^2$  checks!

Global test needed!

First consider the one dimensional case:

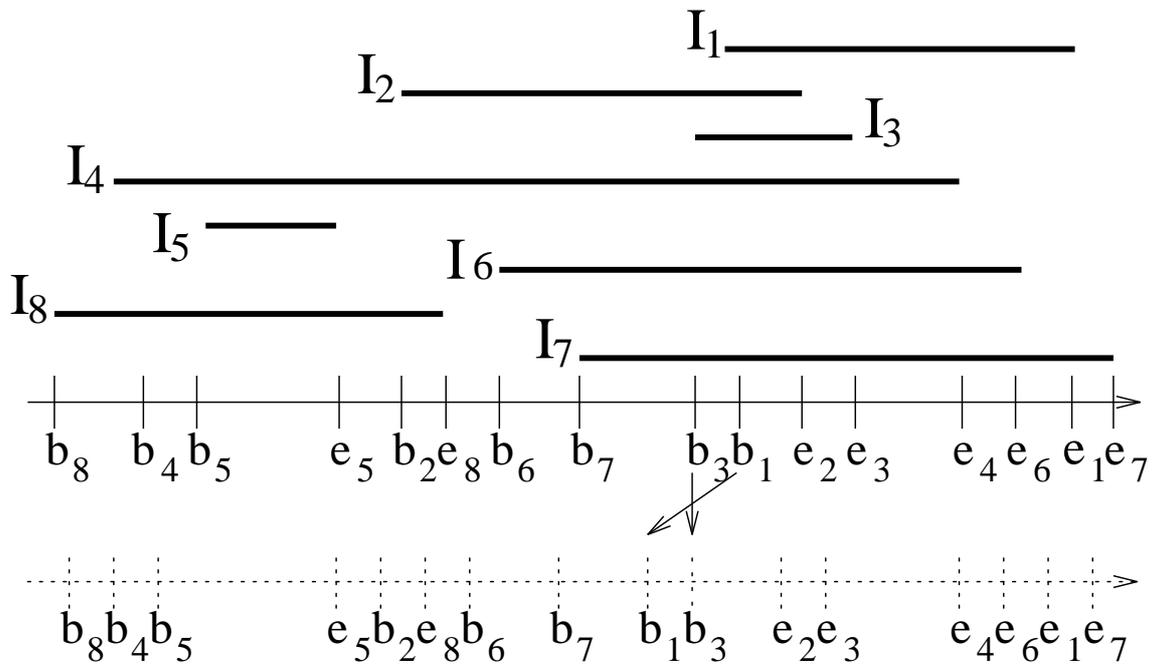


- Sorted List of all boundaries
- Extract collisions from this list with the **sweep** algorithm  $\mathcal{O}(N)$ .
- The best sorting algorithm works with  $\mathcal{O}(N \log N)$ .

$\Rightarrow$  Total cost of  $\mathcal{O}(N \log N)$ .

**Really?**

- $\mathcal{O}(N \log N)$  is the **worst case**
- There are often cases, algorithms are much faster



Reuse the list of the bounding box boundaries sorted in the last time-step.

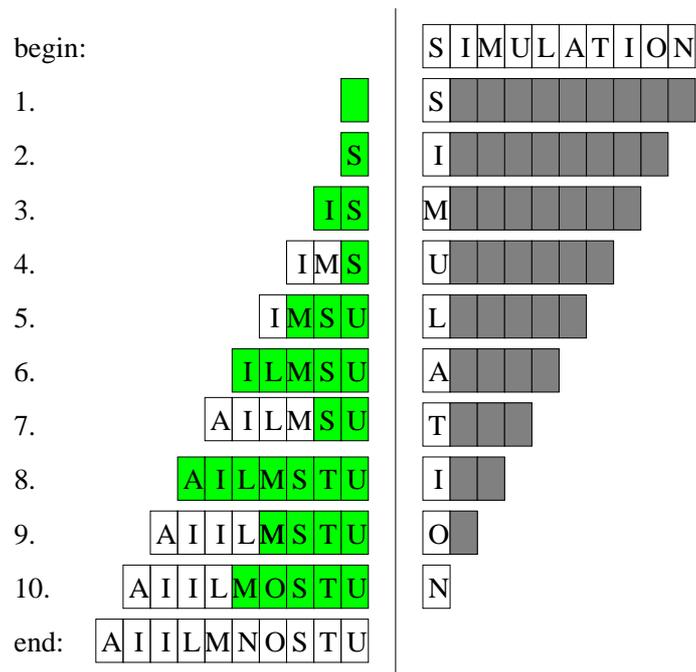


Sort a nearly sorted list.

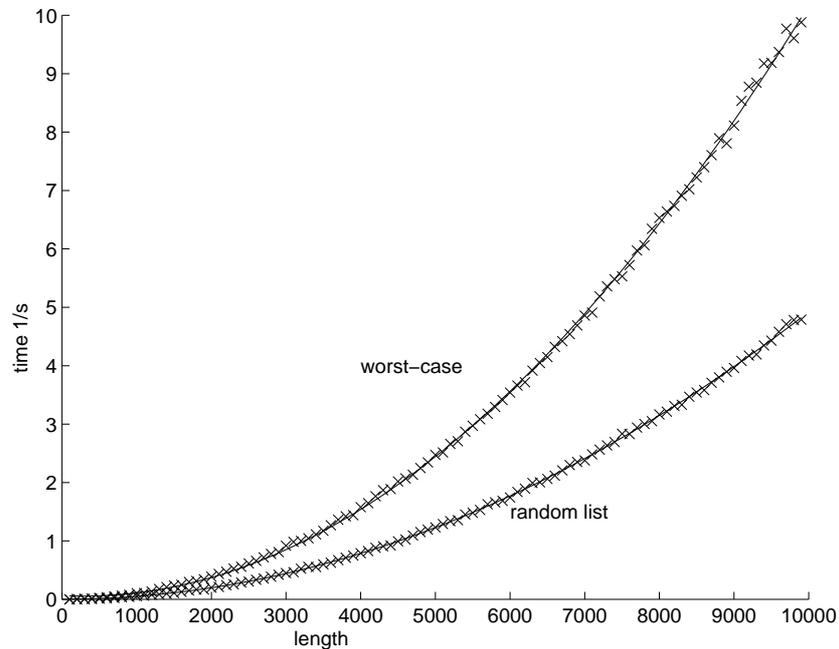


Insertion Sort

# Insertion Sort

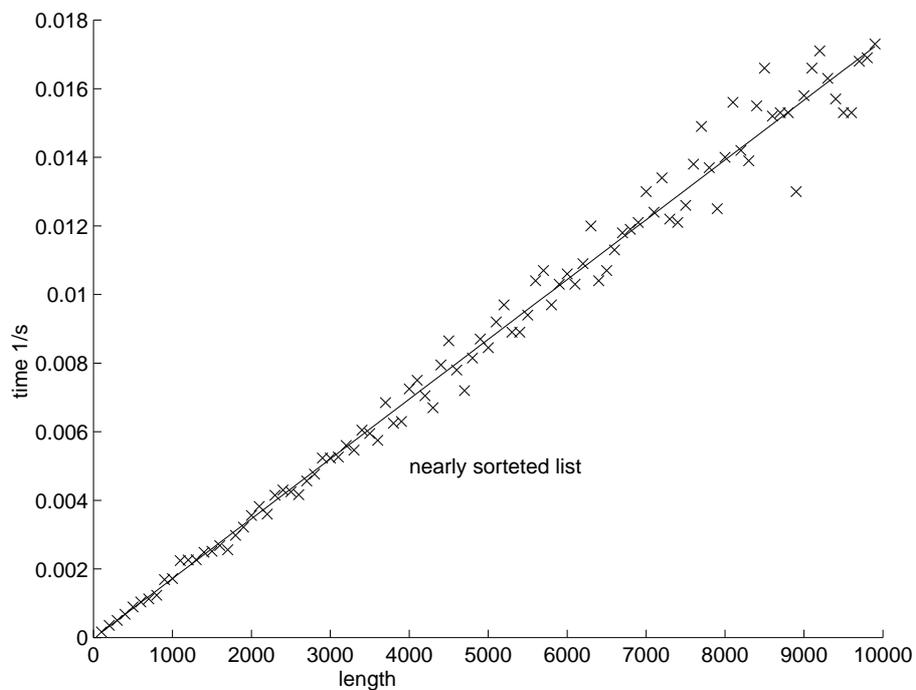


Normally, Insertion sort **normally** works with  $\mathcal{O}(N^2)$ .



## Insertion Sort

- nearly sorted list
- Compare with 1 neighbor
- $\mathcal{O}(N)$  comparisons and nearly no overhead



## Problems

- 2 Steps for finding the overlap on one axis
- in 2 Dimensions get 2 List for X and Y axis
- correlating 2 lists is expensive

Maintain a list of all bounding box collisions:

Want happens, if two boundaries change place?

⇒ Overlap status **may** change

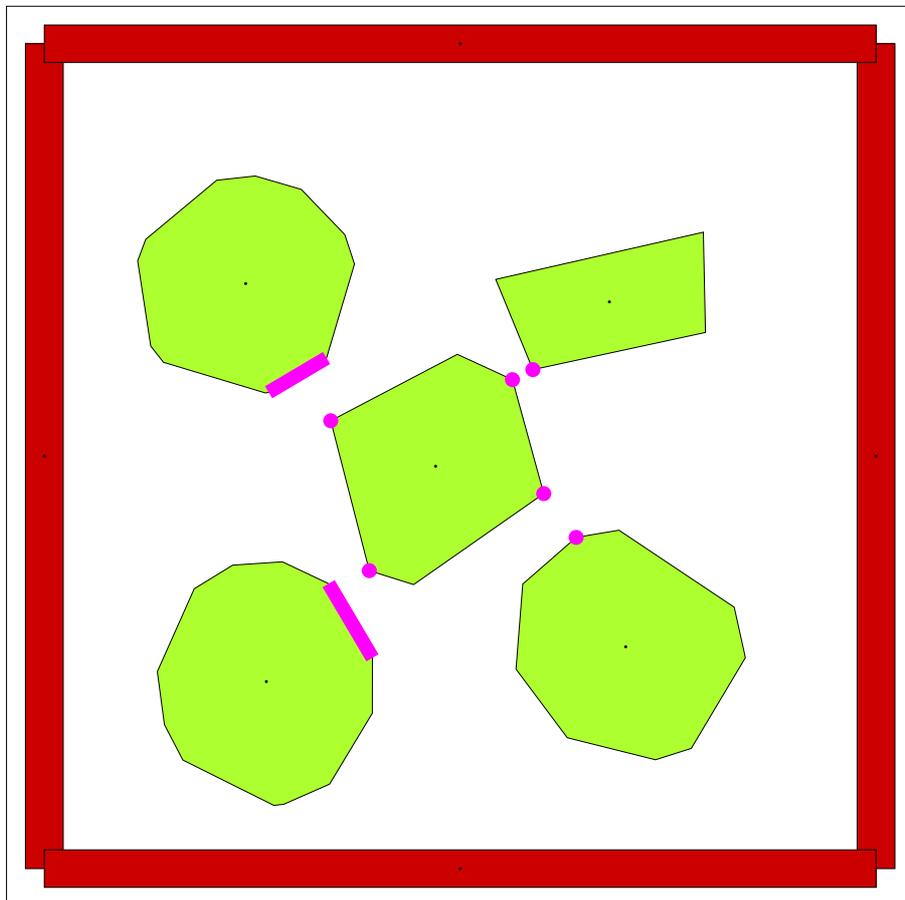
	Y-Axis No Collision	Y-Axis Collision
$bb \rightarrow bb$	$\bar{C} \rightarrow \bar{C}$	$C \rightarrow C$
$ee \rightarrow ee$	$\bar{C} \rightarrow \bar{C}$	$C \rightarrow C$
$be \rightarrow eb$	$\bar{C} \rightarrow \bar{C}$	$C \rightarrow \bar{C}$
$eb \rightarrow be$	$\bar{C} \rightarrow \bar{C}$	$\bar{C} \rightarrow C$

$C = \text{collision}$   $\bar{C} = \text{no collisions}$

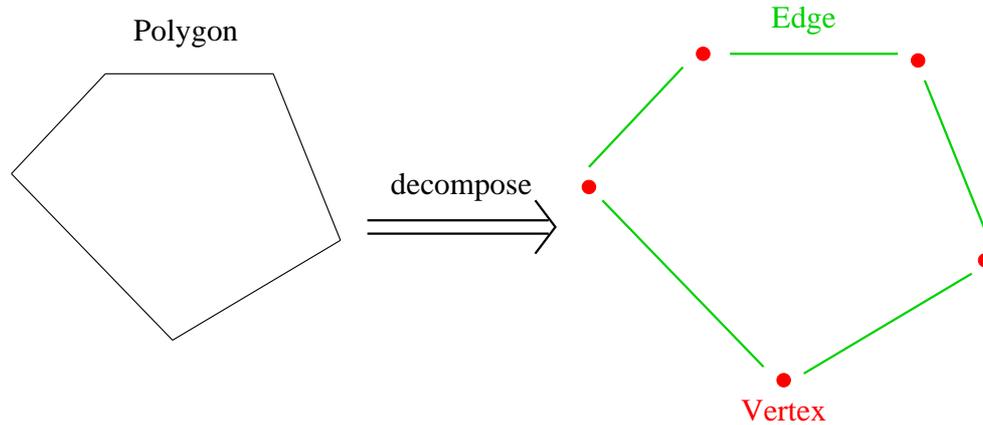
⇒ Update collision list in in one step with  $\mathcal{O}(N)$

## Closest Feature Algorithm

- Calculate distance of particles
- keep track of the closest features
- local test to confirm the distance



## Feature



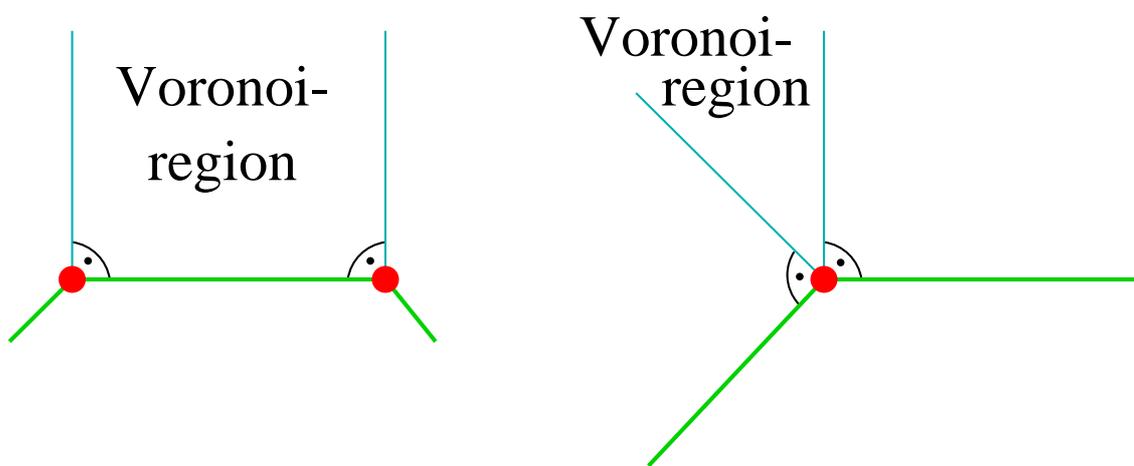
A feature has information on its

- type (edge or vertex)
- geometry
- neighbors
- Voronoi regions
- and more...

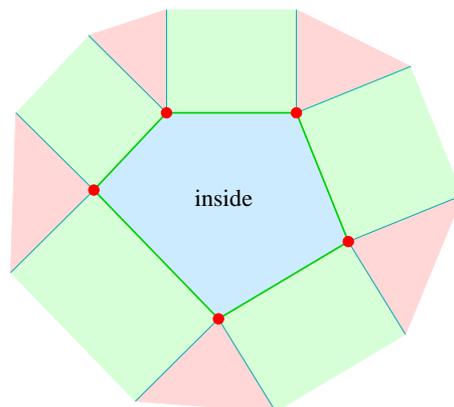
## Voronoi Region

If a point  $p$  lies inside the Voronoi region of a feature  $f_a$ , then

$$\text{dist}(p, f_a) \leq \text{dist}(p, f_b)$$

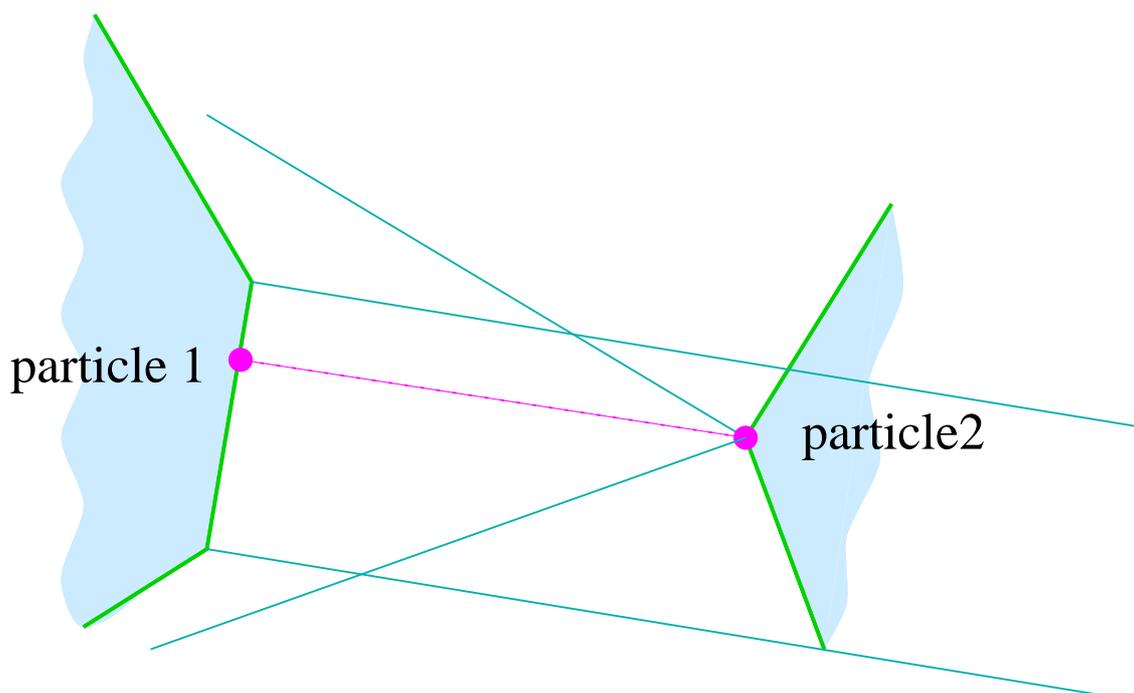


This divides the space **outside** a particle:



## Closest features

If a point  $P$  on object  $P_1$  lies inside the Voronoi region of  $f_2$  on object  $P_2$ , then  $f_2$  is a closest feature to the point  $P$  and vice versa for an Voronoi region of  $f_1$ . If we have a pair of features fulfilling the above condition, we have a pair of closest features.



## Closest Feature Algorithm

We are looking on two features  $f_1$  and  $f_2$  on two polyhedra  $P_1$  and  $P_2$ .

1. Calculate the Voronoi regions  $V_1$  and  $V_2$
2. Calculate a point  $p_1$  on  $f_1$  that is the closest to  $f_2$  and a point  $p_2$  on  $f_2$  that is the closest to  $f_1$
3. Check for  $p_1 \in V_2$ . **If not: choose new  $f_1$  and restart algorithm**
4. Check for  $p_2 \in V_1$ . **If not: choose new  $f_2$  and restart algorithm**

## vertex — vertex

- trivial closest points
- if a Voronoi test fails, continue with neighbor, whose boundary is violated

## vertex — edge

- most frequent case
- special case, if vertex lies **under** the edge
- if a Voronoi test fails, continue with neighbor, whose boundary is violated

## edge — edge

- check for intersection
- if no intersection, replace one edge by vertex

## function check\_vertex\_vertex

```
erg = check_voronoi_of_vertex(vertex1,
                               *vertex2->x1, *vertex2->y1);
switch(erg){
case CF_INSIDE:
    break;
case CF_VIOLATES_PREVIOUS:
    *in_vertex1=(*in_vertex1)->previous;
    return(CF_NO_MINIMUM);
case CF_VIOLATES_NEXT:
    *in_vertex1=(*in_vertex1)->next;
    return(CF_NO_MINIMUM);
}
erg = check_voronoi_of_vertex(vertex2,
                               *vertex1->x1, *vertex1->y1);
switch(erg){
case CF_INSIDE:
    return(CF_MINIMUM);
case CF_VIOLATES_PREVIOUS:
    *in_vertex2=(*in_vertex2)->previous;
    return(CF_NO_MINIMUM);
case CF_VIOLATES_NEXT:
    *in_vertex2=(*in_vertex2)->next;
    return(CF_NO_MINIMUM);
}
```

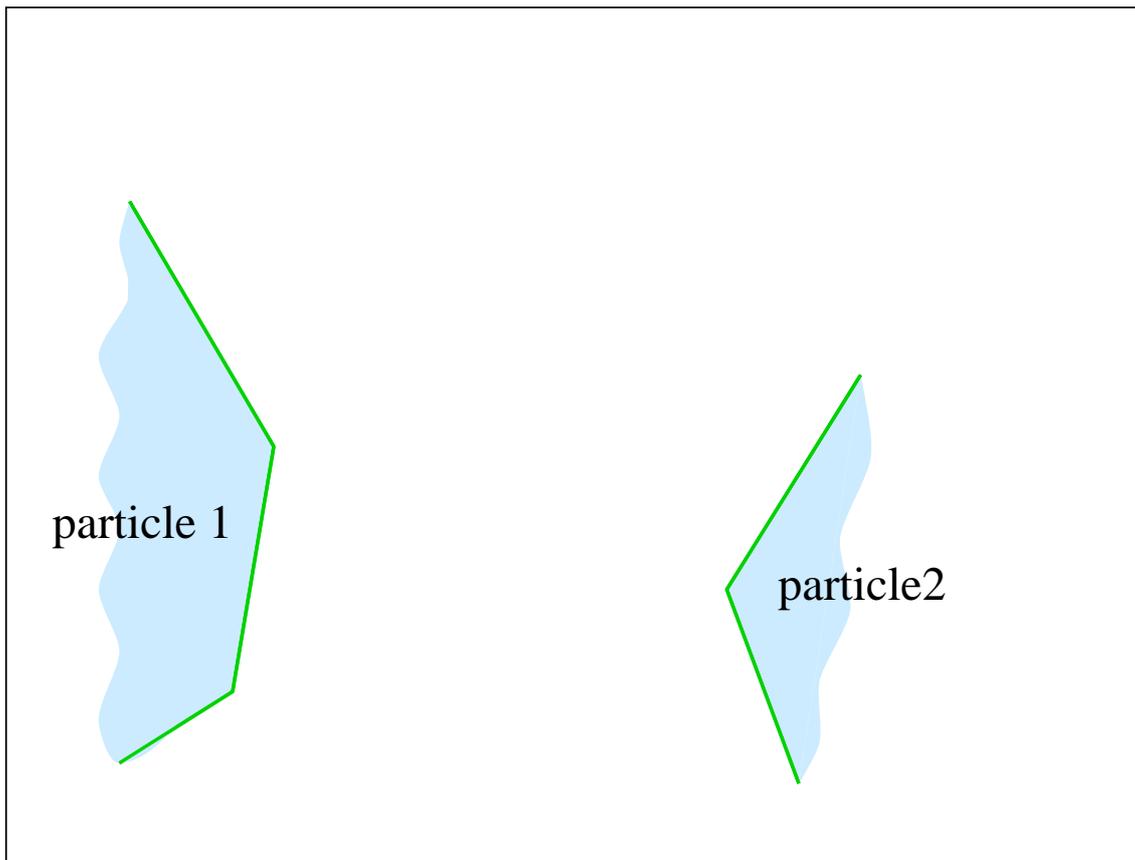
## function find\_closest\_features

```
do{
    type1=coll->feature1->type;
    type2=coll->feature2->type;

    if ((type1==CF_VERTEX)&&(type2==CF_VERTEX)){
        erg=check_vertex_vertex(&(coll->feature1),
            &(coll->feature2),&distance);
    } else if ((type1==CF_VERTEX)&&(type2==CF_EDGE)){
        erg=check_vertex_edge(&(coll->feature1),
            &(coll->feature2),&distance);
    } else if ((type1==CF_EDGE)&&(type2==CF_VERTEX)){
        erg=check_vertex_edge(&(coll->feature2),
            &(coll->feature1),&distance);
    } else {
        erg=check_edge_edge(&(coll->feature2),
            &(coll->feature1),&distance);
    }
}while(erg==CF_NO_MINIMUM);

if (distance<=0){
    coll->colliding=TRUE;
}else{
    coll->colliding=FALSE;
}
```

## Example

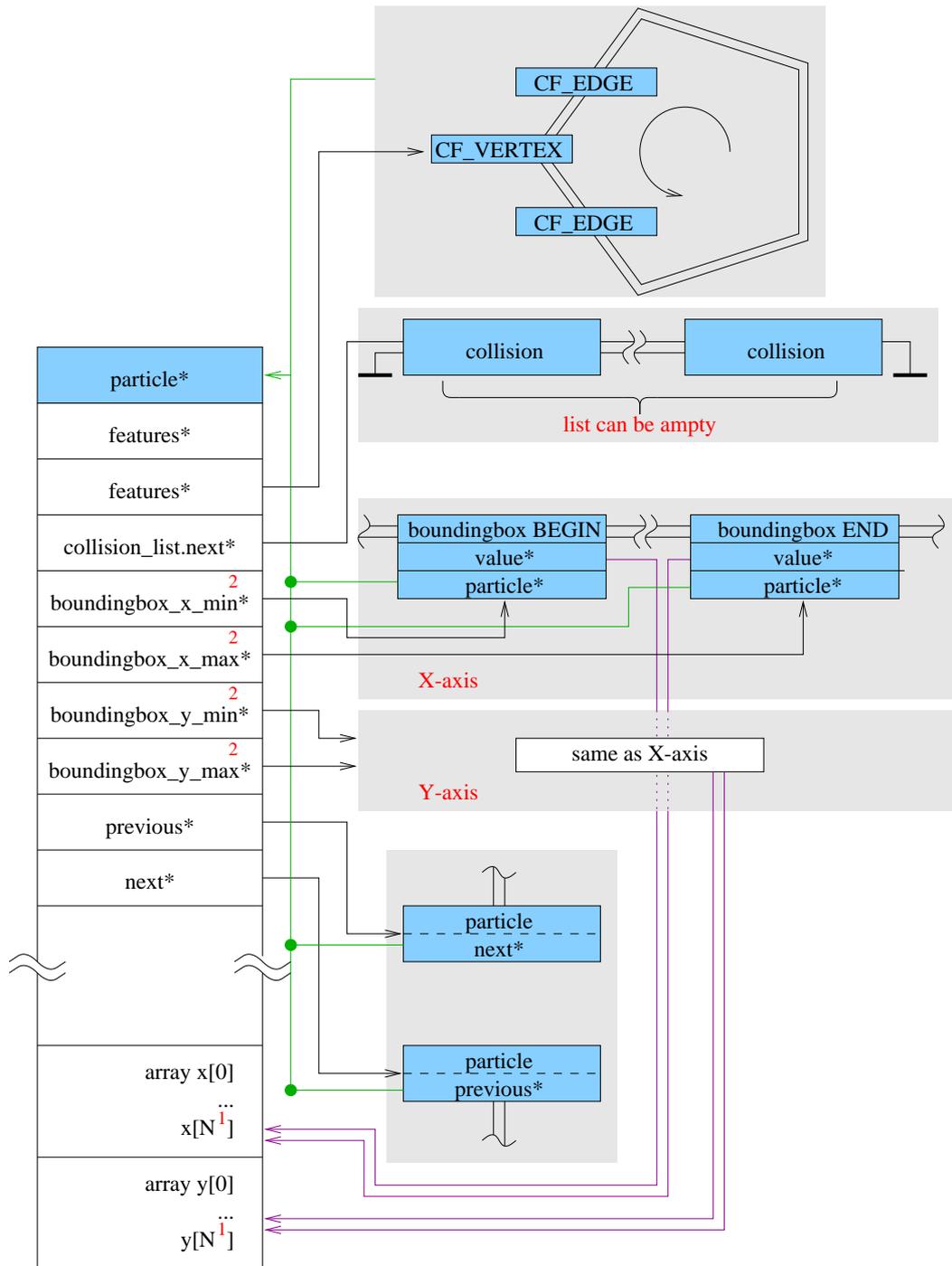


## disadvantages

- complicate program structure
- needs a lot of memory

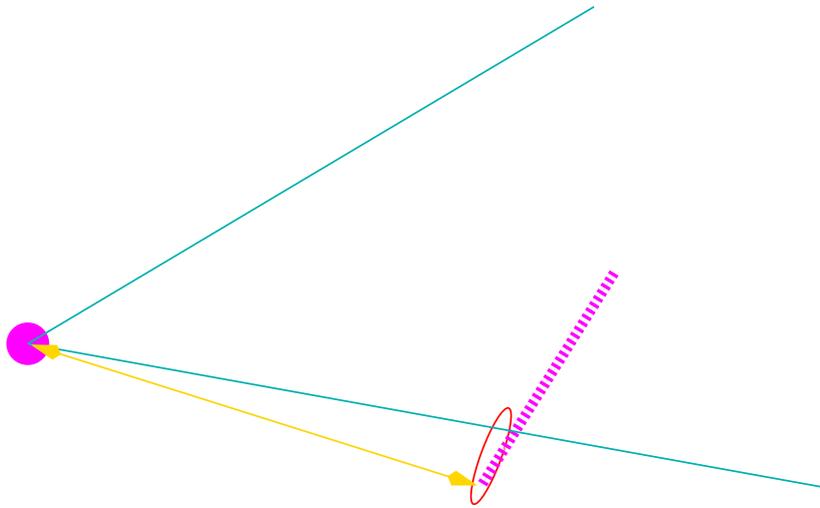
## advantages

- the total cost for the bounding box algorithm is  $\mathcal{O}(N)$
- the total cost for the closest feature algorithm is  $\mathcal{O}(\text{const})$
- also available in 3 dimensions
- only local test, parallel model for shared memory systems possible
- FAST!

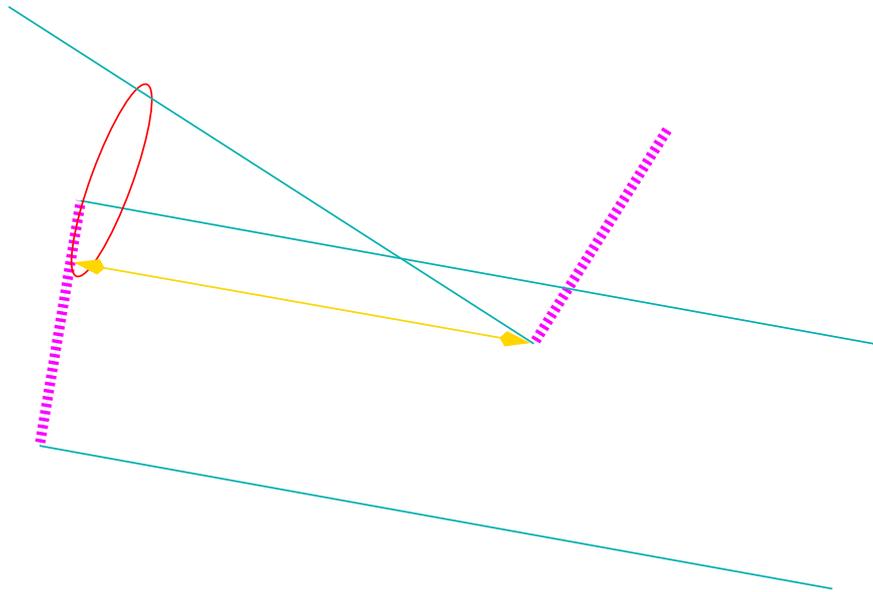


<sup>1</sup> N:=MAX\_NUM\_CORNER  
<sup>2</sup> boundingbox\_\*\_\* are type of SENTINEL

Iteration #1



Iteration #2



Iteration #3

