

Lecture 06 Outline

- * Problem definition and motivations
- * Proximity queries for convex objects
 - → Minkowski space, CSO, Support function
- * GJK based algorithms (GJK, EPA, ISA-GJK)
- Voronoi Clipping Algorithm (V-Clip)
- * Signed Distance Maps for collision detection
- * Demos / tools / libs

Narrow-Phase Collision Detection

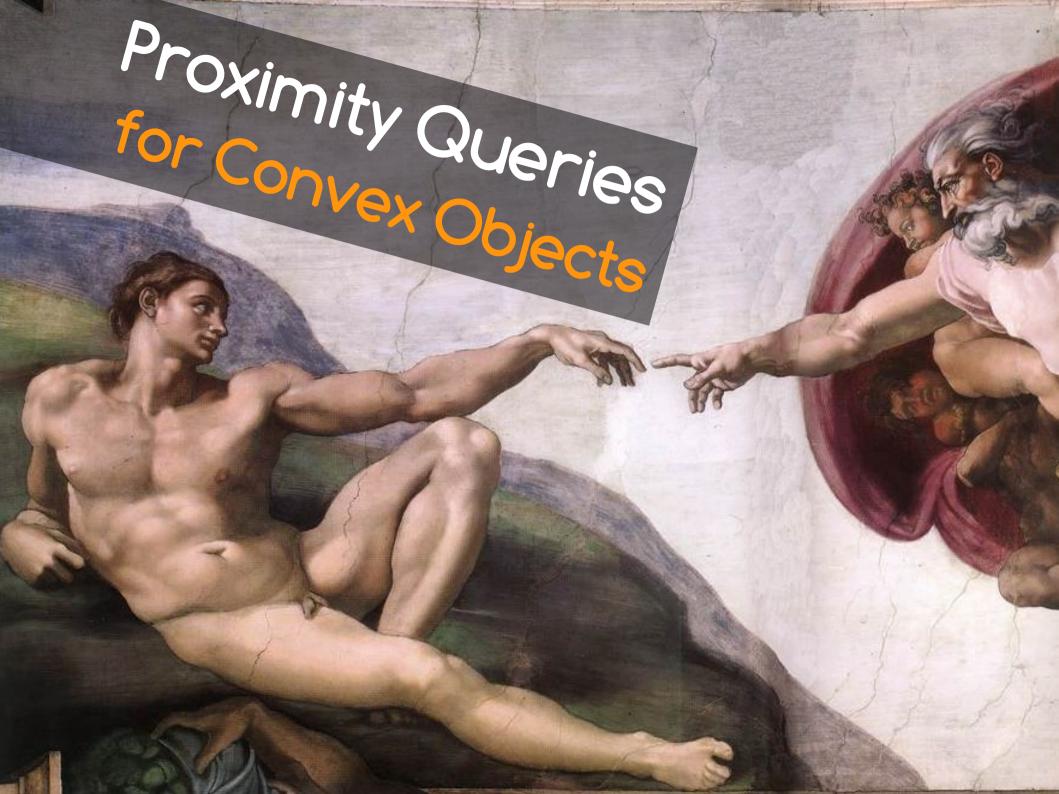
- * Input: List of pairs of potentially colliding objects.
- * Problem1: Find which sub-objects are really intersecting and remove all non-colliding pairs.
- * Problem2: Determine the proximity/contact information, i.e. exact points where objects are touching (interpenetrating), surface normal at that contact point and separating / penetrating distance of objects.
- * Problem3: Recognize persistent contacts, i.e. topologically equivalent contacts from previous time steps

Narrow-Phase Collision Detection

* Output: List of contact regions with necessary proximity information between colliding objects

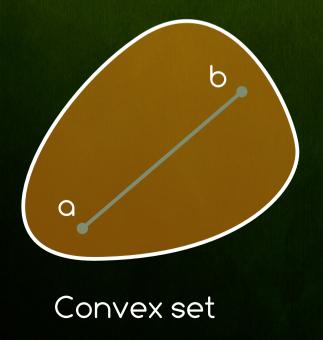
* Strategies:

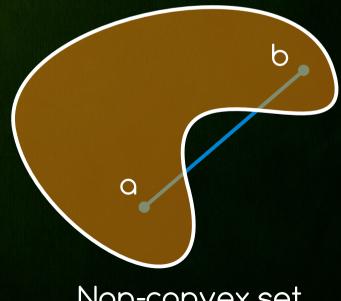
- Simplex based traversal of CSO GJK based algorithms
- Feature tracking base algorithms as Lin-Canny or V-Clip
- Signed Distance Maps for collision detection
- Persistent clustering for contact generation and reduction



Minkowski Space

- * Convex Bounded Point Set
 - ightharpoonup A set S of points $ho \in \mathbb{R}^n$ is called convex and bounded if for any two points a and b the line segment ab lies entirely in S and the distance |a - b| is finite (at most β)
 - \Rightarrow a \in S \land b \in S \land t \in (0, 1) \Rightarrow (1 t)a + tb \in S \land |a b| \leq β
 - > S must be continuous, but needs not to be smooth





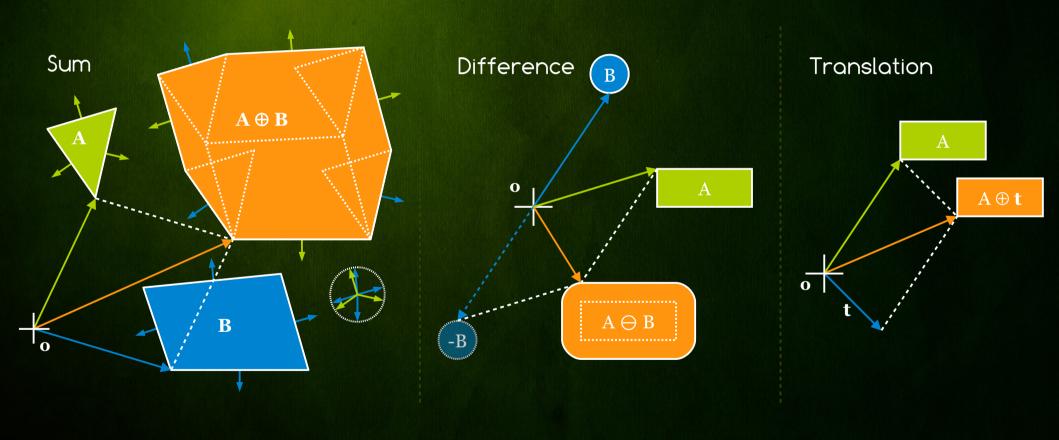
Non-convex set

Minkowski Space

* Given any two convex objects A and B we define Minkowski Sum, Difference and Translation as

- * Minkowski Sum A ⊕ B
 - \rightarrow A \oplus B = {a + b | a \in A \land b \in B}
- * Minkowski Difference A ⊖ B (known as CSO)
 - \rightarrow A \ominus B = A \oplus (-B) = {a b | a \in A \land b \in B}
- - \rightarrow A \oplus t = A \oplus {t} = {a + t | a \in A}

Minkowski Space



Touching Vectors

* Touching Contact

Two convex objects A and B are in touching contact, iff their intersection (as a point set) is a subset of some (contact) plane β . Formally: A \cap B \subset β

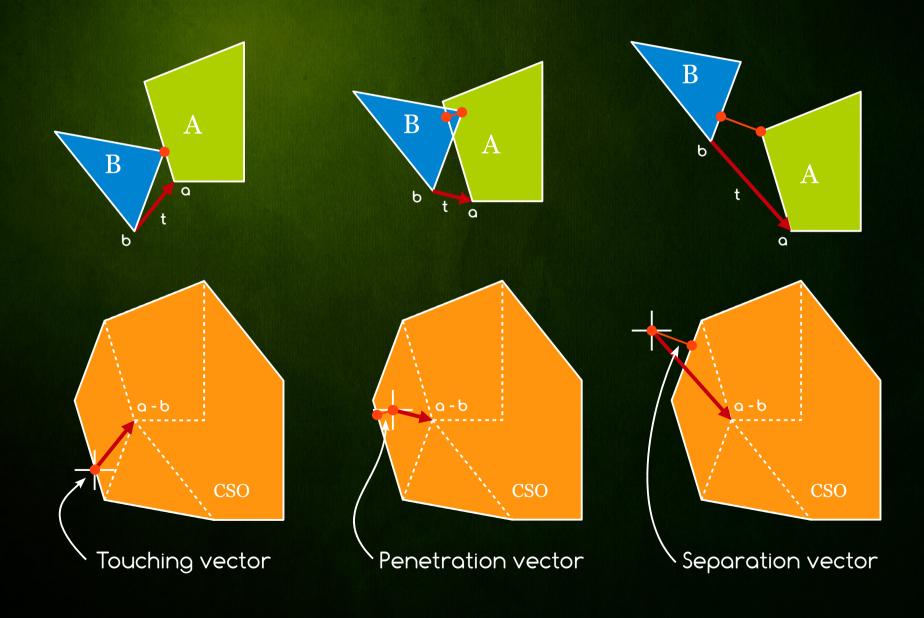
* Touching Vector

- → The touching vector t_{AB} between two convex objects A and B is any shortest translational vector t moving objects into the touching contact.
- ullet ullet

* Touching Distance

- > Touching distance d_{AB} is the length of touching vector t_{AB}.
- \rightarrow d_{AB} = min {|t| | A \cap (B \oplus t) \subset β \wedge t \in R³ }

Touching Vectors and CSO



Touching Vectors

- * Objects are in close proximity if their touching distance is smaller than a defined threshold
- * If objects are disjoint touching vector (distance) is usually called as separation vector (distance)
- * If objects are intersecting touching vector (distance) is usually called as penetration vector (depth)
- * Separation vector is unique. Penetration vector is usually not unique (co-centric circles)

Support Set and Boundary

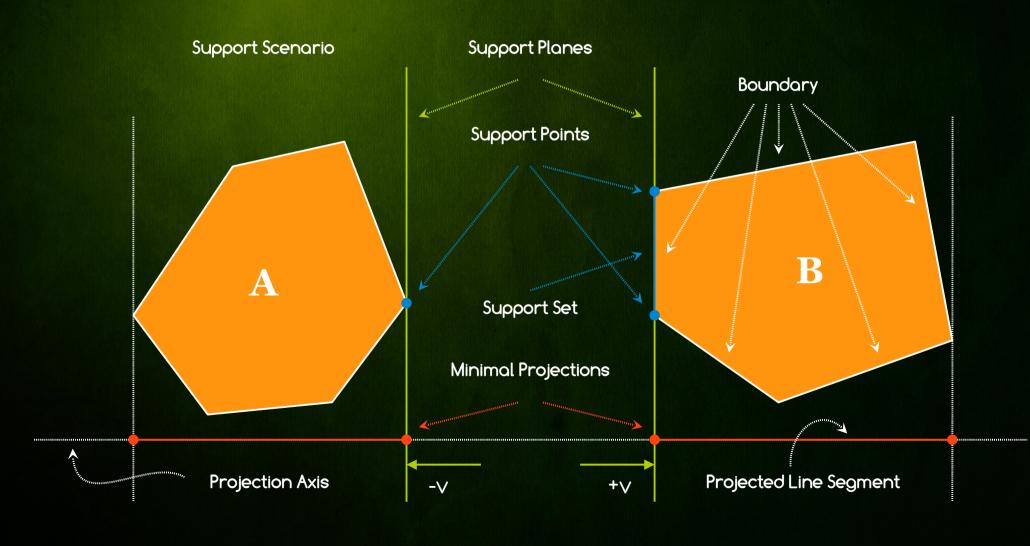
* Support Set

- The set of points from a convex object C which have a minimal projection onto a direction axis d is the support set of C
- $\Rightarrow S_C^d = \{ \rho \mid \rho \in C \land d^T \rho = \min \{ d^T c \mid c \in C \} \}$

* Support Boundary

- The set of all support points from a convex object C with respect to any direction d is the boundary of C
- \rightarrow $\partial(C) = \{ \rho \mid \rho \in S_C^{d} \land d \in R^3 \}$

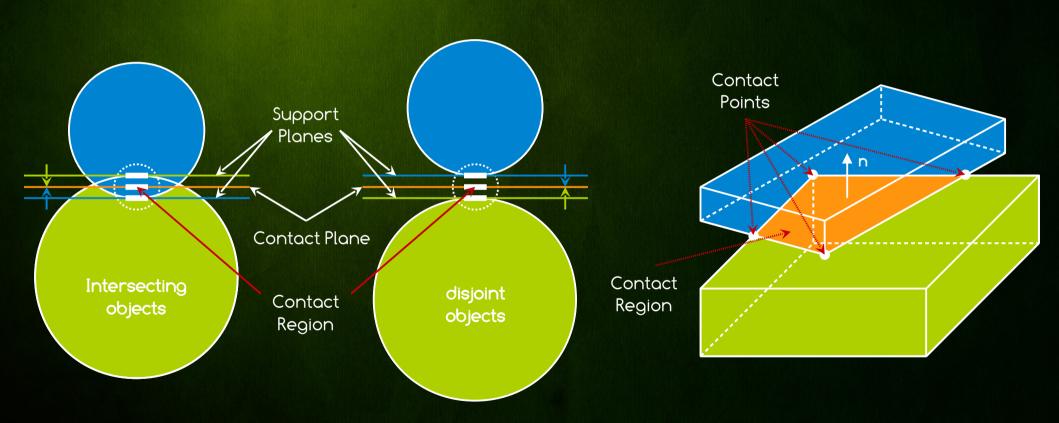
Support Set and Boundary



Touching Vectors and Boundary

- * Touching Vector Theorem
 - Any translational vector t moves two convex objects A and B into touching contact, iff it lies on the boundary of their CSO
 - \rightarrow A \cap (B \oplus t) \subset $\beta \Leftrightarrow$ t \in ∂ (A \ominus B)
- * This theorem can simplify the definition of touching contact, vector and distance, by replacing (A \cap (B \oplus t) \subset β) with the t \in ∂ (A \ominus B)
 - → d_{AB} = min { $|t| | t \in \partial(A \ominus B)$ }
 - ightharpoonup $t_{AB} \in \{ t \mid t \in \delta(A \ominus B) \land |t| = d_{AB} \}$

Contact Region



Contact Region

- * If objects are in touching contact (t_{AB} is zero), their intersection simply forms the contact region
- * If objects penetrate or are disjoint (t_{AB} is non-zero) contact region is constructed as follows
 - ightharpoonup Compute two support sets S_A^{+tAB} and S_B^{-tAB} for A and B w.r.t \mathbf{t}_{AB}
 - ightharpoonup Project both sets onto touching vector $\mathbf{t}_{\scriptscriptstyle{AB}}$ and take median
 - ightharpoonup Form contact plane with median as origin and normal as $\mathbf{t}_{\scriptscriptstyle{\mathrm{AB}}}$
 - Project both support sets onto contact plane and take their (ideally) intersection as contact region



Gilbert - Johnson - Keerthi Algorithm

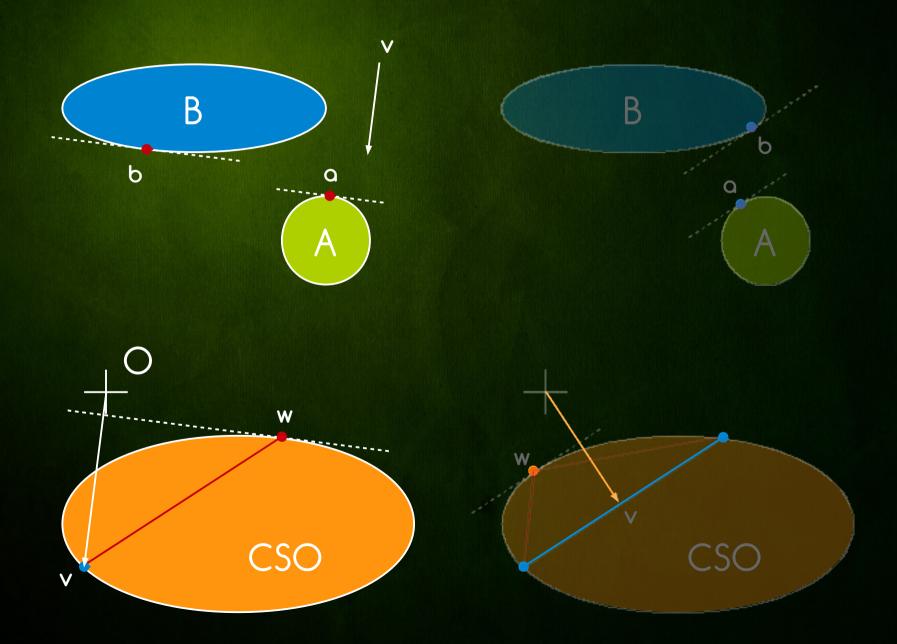
Gilbert - Johnson - Keerthi Algorithm

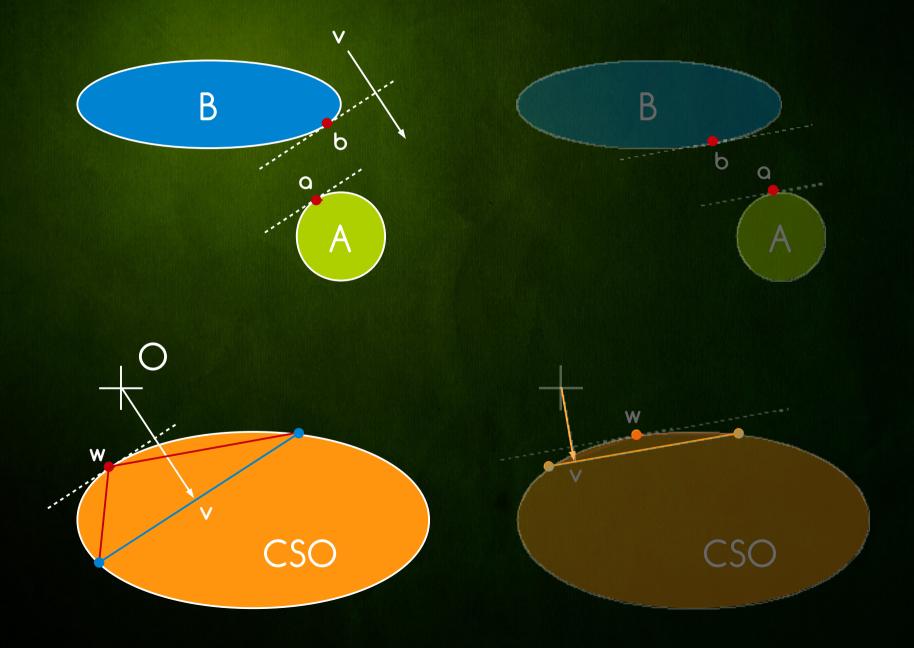
- * Key idea of all GJK based algorithms: iterative search for the touching vector in CSO
- * Strategy: Perform a descent traversal of the CSO surface to find the closest point to the origin
- * Problem: Naive construction and traversal of CSO is expensive and slow
- * Solution: Simple support function can select proper support points on CSO and thus speed up the traversal to an almost constant time assuming coherent simulation.

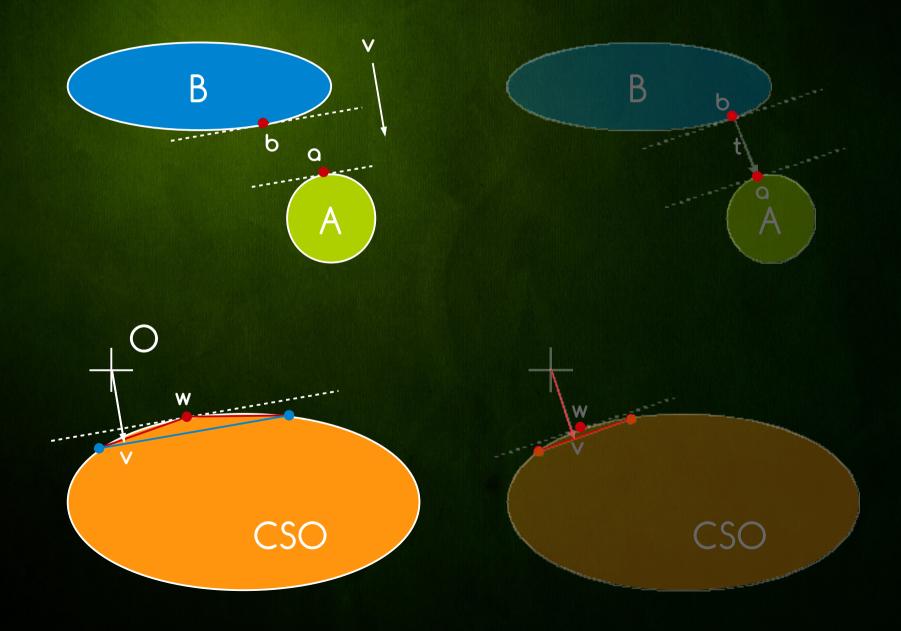
Support Function

- * Support function support (C,d) \in S^d_c of a convex object C w.r.t. direction d simply returns any support point from the respective support set S^d_c
- Support Function Operations
 - Assuming support (A, d) \in S^d_A and support (B, d) \in S^d_B, we define the support functions as follows
 - \rightarrow support(-B, d) = -support(B,-d) $\in S_{-B}^{d}$
 - \rightarrow support(A \oplus B, d) = support(A, d) + support(B, d) \in S^d_{A \oplus B}
 - ⇒ support($A \ominus B$, d) = support($A \ominus (-B)$, d) = support(A, d) + support(A, d) = support(A,+d) - support(A,-d)

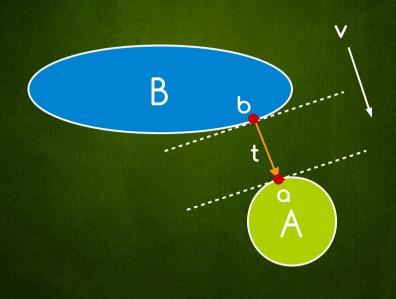
- * The traversal is done by iteratively constructing a sequence of simplices in 3D
 - point or line or triangle or tetrahedron
- * In each iteration newly created simplex is closer to the origin as the one in previous iteration
- * New simplex is created by
- * 1) Adding a support point to the former simplex
- * 2) Taking the smallest sub-simplex which contains the closest point to the origin

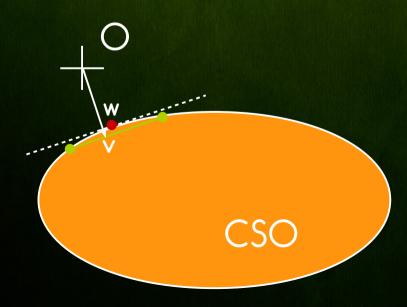






Proximity GJK Algorithm Algorithm





```
In: Convex objects A, B and initial simplex W
Out: Touching vector w
function ProximityGJK(A, B, W): w
      \{\mathbf{v},\ \delta\} \leftarrow \{\mathbf{1},\ 0\}
        while (\|\mathbf{v}\|^2 - \delta^2 > \varepsilon) do
 2:
              \mathbf{v} \leftarrow \text{ClosestPoint}(W)
 3:
              \mathbf{w} \leftarrow \text{SUPPORT}(A \ominus B, \mathbf{v}) = \text{SUPPORT}(A, +\mathbf{v}) - \text{SUPPORT}(B, -\mathbf{v})
4:
             W \leftarrow \text{BestSimplex}(W, \mathbf{w})
 5:
              if (|W| = 4) then return ProximityEPA(A, B, W);
6:
              if (\mathbf{v}^{\mathrm{T}}\mathbf{w} > 0) then \delta^2 \leftarrow \max \left\{ \delta^2, \frac{(\mathbf{v}^{\mathrm{T}}\mathbf{w})^2}{\|\mathbf{v}\|^2} \right\}
7:
        end
8:
        return w
 9:
end
```

Computing Support Function

- * Searching for the support vertex w heavily depends on the representation of the convex objects A and B
- * For a simple primitives it can be computed directly
- * For convex polytopes
 - Naive approach is to project all vertices onto the direction axis and take any one with the minimal projection
 - → if we consider a coherent simulation we can use a local search sometimes called as "hill climbing" and find the support vertex in almost constant time

Hill Climbing Support Function

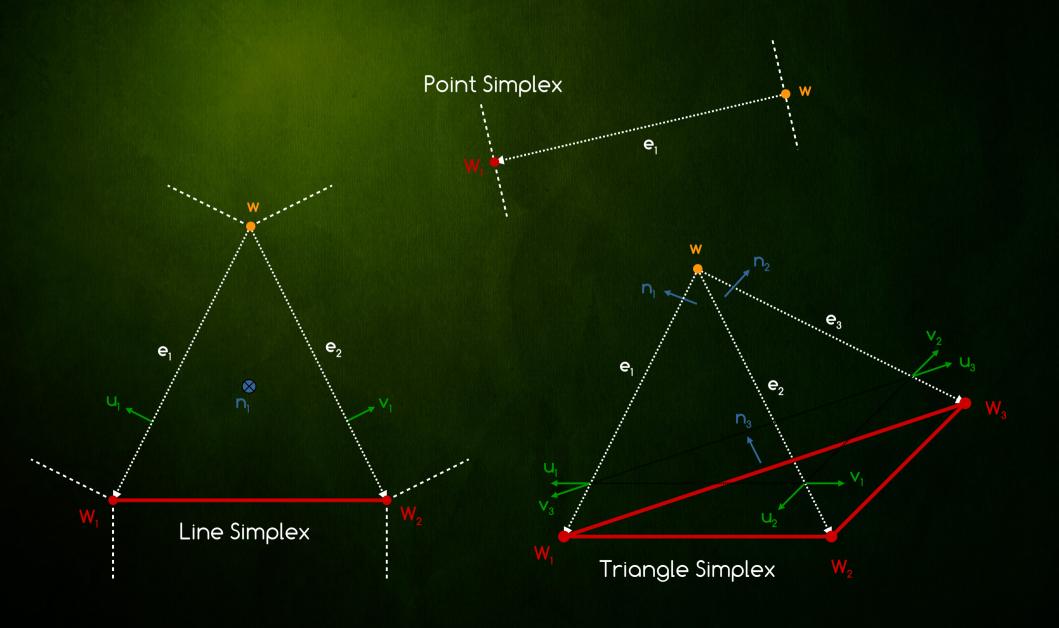
*For convex polytopes do a local search to "refine" the support point from previous simulation state

```
In: Convex polytope A, initial support vertex \mathbf{w} and the direction vector \mathbf{d}
Out: New support vertex with minimal projection w
function SupportHC(A, \mathbf{d}, \mathbf{w}): w
       \{\mu, \text{ Found}\} \leftarrow \{\mathbf{d}^{\mathrm{T}}\mathbf{w}, \text{ false}\}
       while not Found do
2:
            Found \leftarrow true
3:
            foreach w' in Neighbours(w) do
4:
                  if (\mathbf{d}^T \mathbf{w}' < \mu) then \{\mu, \mathbf{w}, \text{ Found}\} \leftarrow \{\mathbf{d}^T \mathbf{w}', \mathbf{w}', \text{ false}\}; break
5:
            end
6:
       end
7:
       return w
8:
end
```

Simplex Refinement

- * Problem: Given a simplex and new vertex form new simple by adding the vertex and select subsimplex closest to the origin
- * Bad solution: The simplex can be done by solving a system of linear equations (slow, numeric issues)
- * Good solution: Form new simplex and test in which external Voronoi region the origin lies.
- * The selected Voronoi region directly shows us which sub-simplex is the desired (closest) one

Voronoi Simplex Refinement



Voronoi Simplex Refinement

- * Empty Simplex: A vertex simplex {w} is formed
 - → The smallest simplex, which contains the closest point to the origin is {w} (case 0)
- * Vertex Simplex: An edge simplex {W1,w} is formed
 - → It has 2 vertex regions {W1, w} and one edge region {e1}
 - Since W1 lies on support plane which is perpendicular to the support axis (vector w) origin can not be in the region of W1
 - Thus we check only regions of w and e1 by projecting -w onto the edge e1 (case 1)

Voronoi Simplex Refinement

- * Edge Simplex: A face simplex {W1,W2,w} is formed
 - → It has 3 vertex regions, 3 edge regions and 2 face regions
 - The origin can be only in {w, e1, e2, n1} regions
 - Construct Voronoi planes with normals {e1, e2, u1, v1} and test whether the origin is above or below these planes, i.e. compare signs of -w projections onto these normals
- *Face Simplex: A tetrahedron simplex {W1,W2,W3, w} is formed
 - A tetrahedron has 4 vertex regions, 4 face regions, 6 edge regions and 1 interior region (T)
 - Origin can lie only only in regions {w, e1, e2, e3, n1, n2, n3,T}
 - Construct Voronoi planes with normals {e1, e2, e3, n1, n2, n3, u1, u2, u3, v1, v2, v3} and test sign -w projection onto normals

In: Simplex W and new point on CSO surface \mathbf{w} Out: New smallest simplex W containing \mathbf{w} and the closest point to the origin function BestSimplex $(W, \mathbf{w}) : W$ $d \leftarrow 0 - w$ 2: $\mathbf{e}_1 \leftarrow \mathbf{W}_1 - \mathbf{w}$: $\mathbf{e}_2 \leftarrow \mathbf{W}_2 - \mathbf{w}$: $\mathbf{e}_3 \leftarrow \mathbf{W}_3 - \mathbf{w}$: 3: $\mathbf{n}_1 \leftarrow \mathbf{e}_1 \times \mathbf{e}_2$: $\mathbf{n}_2 \leftarrow \mathbf{e}_2 \times \mathbf{e}_3$: $\mathbf{n}_3 \leftarrow \mathbf{e}_3 \times \mathbf{e}_1$: 4: $\mathbf{u}_1 \leftarrow \mathbf{e}_1 \times \mathbf{n}_1$: $\mathbf{u}_2 \leftarrow \mathbf{e}_2 \times \mathbf{n}_2$: $\mathbf{u}_3 \leftarrow \mathbf{e}_3 \times \mathbf{n}_3$; 5: $\mathbf{v}_1 \leftarrow \mathbf{n}_1 \times \mathbf{e}_2$: $\mathbf{v}_2 \leftarrow \mathbf{n}_2 \times \mathbf{e}_3$: $\mathbf{v}_3 \leftarrow \mathbf{n}_3 \times \mathbf{e}_1$: switch |W| do 6: case 0 /* empty simplex */ 7: return {w} 8: 9: end /* vertex simplex */ case 1 10: if $(\mathbf{d}^{\mathrm{T}}\mathbf{e}_{1} > 0)$ then return $\{\mathbf{w}\}$ 11: if $(\mathbf{d}^{\mathrm{T}}\mathbf{e}_{1} < 0)$ then return $\{\mathbf{W}_{1}, \mathbf{w}\}$ 12: end 13: case 2 /* edge simplex */ 14: if $(\mathbf{d}^{\mathrm{T}}\mathbf{e}_{1} < 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{e}_{2} < 0)$ then return $\{\mathbf{w}\}$ 15: if $(\mathbf{d}^{\mathrm{T}}\mathbf{e}_1 > 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{u}_1 > 0)$ then return $\{\mathbf{W}_1, \mathbf{w}\}$ 16: if $(\mathbf{d}^{\mathrm{T}}\mathbf{e}_{2} > 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{v}_{1} > 0)$ then return $\{\mathbf{W}_{2}, \mathbf{w}\}$ 17: if $(\mathbf{d}^{\mathrm{T}}\mathbf{u}_{1} < 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{v}_{1} < 0)$ then return $\{\mathbf{W}_{1}, \mathbf{W}_{2}, \mathbf{w}\}$ 18: 19: end case 3 /* face simplex */ 20: if $(\mathbf{d}^T \mathbf{e}_1 < 0) \wedge (\mathbf{d}^T \mathbf{e}_2 < 0) \wedge (\mathbf{d}^T \mathbf{e}_3 < 0)$ then return $\{\mathbf{w}\}$ 21: if $(\mathbf{d}^{\mathrm{T}}\mathbf{e}_1 > 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{u}_1 > 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{v}_3 > 0)$ then return $\{\mathbf{W}_1, \mathbf{w}\}$ 22: $\mathbf{if}\ (\mathbf{d}^T\mathbf{e}_2>0)\quad \wedge\quad (\mathbf{d}^T\mathbf{u}_2>0)\quad \wedge\quad (\mathbf{d}^T\mathbf{v}_1>0)\ \mathbf{then}\ \ \mathbf{return}\ \{\mathbf{W}_2,\mathbf{w}\}$ 23: $\textbf{if} \ (\mathbf{d}^T\mathbf{e}_3 > 0) \quad \wedge \quad (\mathbf{d}^T\mathbf{u}_3 > 0) \quad \wedge \quad (\mathbf{d}^T\mathbf{v}_2 > 0) \ \textbf{then} \ \ \textbf{return} \ \{\mathbf{W}_3, \mathbf{w}\}$ 24: if $(\mathbf{d}^{\mathrm{T}}\mathbf{n}_1 > 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{u}_1 < 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{v}_1 < 0)$ then return $\{\mathbf{W}_1, \mathbf{W}_2, \mathbf{w}\}$ 25: if $(\mathbf{d}^T \mathbf{n}_2 > 0) \land (\mathbf{d}^T \mathbf{u}_2 < 0) \land (\mathbf{d}^T \mathbf{v}_2 < 0)$ then return $\{\mathbf{W}_2, \mathbf{W}_3, \mathbf{w}\}$ 26: if $(\mathbf{d}^{\mathrm{T}}\mathbf{n}_3 > 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{u}_3 < 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{v}_3 < 0)$ then return $\{\mathbf{W}_3, \mathbf{W}_1, \mathbf{w}\}$ 27: if $(\mathbf{d}^{\mathrm{T}}\mathbf{n}_1 < 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{n}_2 < 0) \wedge (\mathbf{d}^{\mathrm{T}}\mathbf{n}_3 < 0)$ then return $\{\mathbf{W}_1, \mathbf{W}_2, \mathbf{W}_3, \mathbf{w}\}$ 28: 29: end 30: end end

Closest Point on Simplex

- *Problem: Given (0 or 1 or 2 or 3) simplex {W1,W2,W3} find the closest point to the origin
- * Empty Simplex: Return 0
- Vertex Simplex: Return W1
- * Edge Simplex: Return the closest point on line {W1,W2} to the origin.
 - → No need to check other regions (eg. vertex W1 region etc.)
- *Face Simplex: Return the closest point on plane {W1,W2,W3} to the origin.
 - No need to check other regions (eg. vertex W1 region etc.)

Closest Point Algorithm

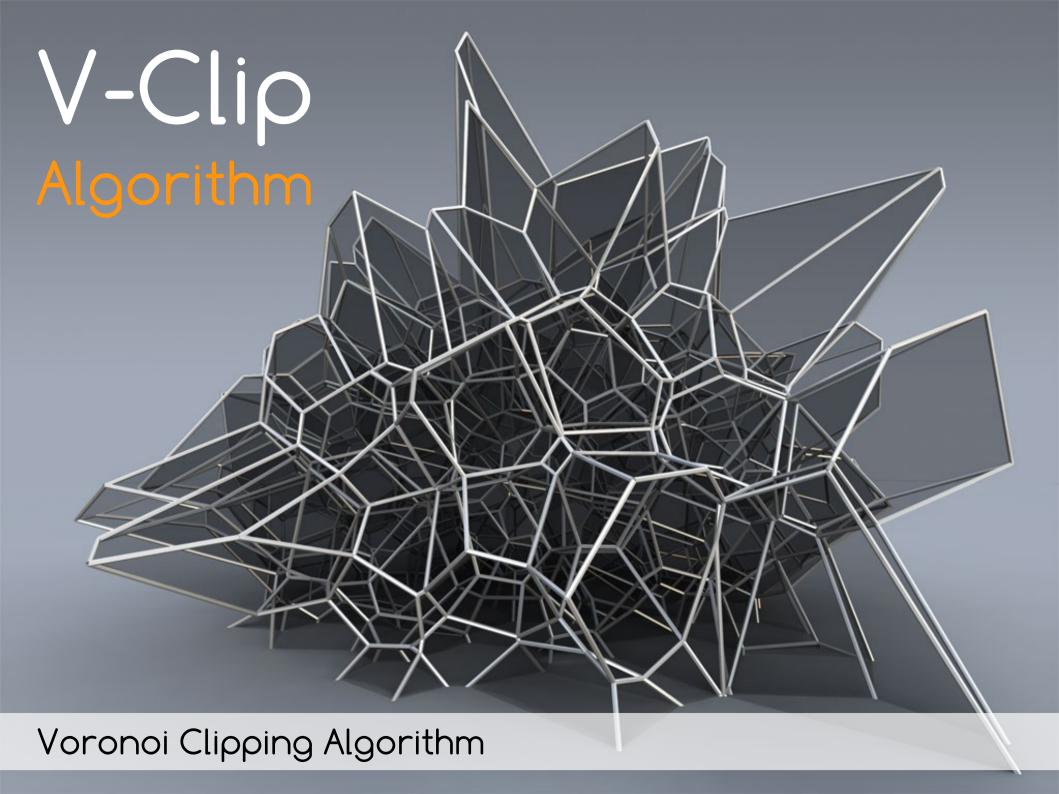
```
In: Simplex W
Out: Closest point on simplex to the origin v
function ClosestPoint(W): \mathbf{v}
      \mathbf{d} \leftarrow \mathbf{W}_2 - \mathbf{W}_1
 1:
       \mathbf{n} \leftarrow (\mathbf{W}_2 - \mathbf{W}_1) \times (\mathbf{W}_3 - \mathbf{W}_1)
 2:
        switch |W| do
 3:
             case \theta return 0;
                                                               /* empty simplex */
4:
                                                             /* vertex simplex */
             case 1 return W_1;
 5:
             case 2 return \mathbf{W}_1 - \frac{\mathbf{d}^T \mathbf{W}_1}{\mathbf{d}^T \mathbf{d}} \mathbf{d}; /* edge simplex */
 6:
             case 3 return \frac{\mathbf{n}^{\mathrm{T}}\mathbf{W}_{1}}{\mathbf{n}^{\mathrm{T}}\mathbf{n}}\mathbf{n};
                                                     /* face simplex */
 7:
        end
 8:
end
```

GJK Overlap Test

- * Incremental Separating-Axis GJK (ISA-GJK)
 - A subtle modification to the proximity GJK
 - Descent overlap test for convex objects
 - Iteratively searches for some separating axis
 - Average constant time complexity in coherent simulation
- * Principle: Similar traversal to Proximity GJK
 - Reports overlap: When the best simplex is tetrahedron
 - Reports no-overlap: When the signed distance of the support plane to the origin is positive
 - → $v^Tw = v^T$ support $(A \ominus B, v) = v^T$ support $(A,+v) v^T$ support (B,-v) > 0

ISA-GJK Algorithm

```
In: Convex objects A, B and initial Simplex W
Out: Overlap check: (true/false)
function OverlapGJK(A, B, W): bool
      \{v, w\} \leftarrow \{1, 1\}
      while (\mathbf{v}^{\mathrm{T}}\mathbf{w} \leq 0) do
2:
           \mathbf{v} \leftarrow \text{ClosestPoint}(W)
3:
           \mathbf{w} \leftarrow \text{SUPPORT}(A \ominus B, \mathbf{v}) = \text{SUPPORT}(A, +\mathbf{v}) - \text{SUPPORT}(B, -\mathbf{v})
4:
           W \leftarrow \text{BestSimplex}(W, \mathbf{w})
5:
           if (|W| = 4) then return true;
                                                                      /* intersection */
6:
       end
7:
       return false
8:
end
```



* Interior Set:

The set of all interior points int (C) of a convex polytope C is the intersection of negative half-spaces formed by all faces of C (surface points are not included)

* int(C) = {
$$c \in R3 \mid ds(c, F) < 0 \land F \in C$$
 }

* Distance:

→ The distance d(c,X) between a feature X and some point c is the minimum distance between c and any point of X

*
$$d(c,X) = min \{ |x-c| | x \in X \}$$

- * Signed Distance
 - → The signed distance $d_s(c, F)$ between a point c and a plane F, defined by a unit normal n_F and a reference point o_F is the projection of the reference vector $(c o_F)$ onto planes normal
- * $ds(c, F) = n_F^T (c o_F)$
- *Having two incident features X, Y: if X has a lower dimension than Y, then X must be a subset of Y and therefore the distance of any point c to X is less than or equal to Y
- * $X \cap Y \wedge dim(X) < dim(Y) \Rightarrow X \subset Y \Rightarrow d(c,X) \leq d(c,Y)$

* External Voronoi Region

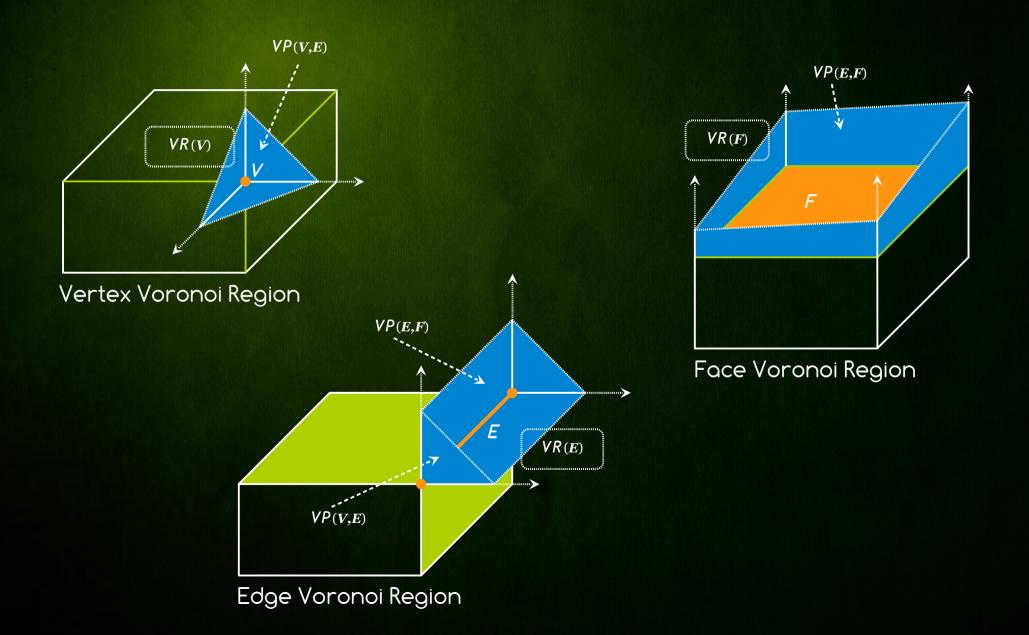
- The Voronoi region VR(X) of a feature X on some convex polytope C is a set of external points which are closer (≤) to X than to any other feature Y in C
- → $VR(X) = \{ c \notin int(C) \mid d(c,X) \leq d(c,Y) \land Y \in C \}$

* External Voronoi Plane

- The Voronoi plane VP(X,Y) of two incident features X and Y is the plane containing the intersection of their Voronoi regions.
- → VP(X,Y) = β \wedge VR(X) \cap VR(Y) \subset β

* Inter-feature Distance

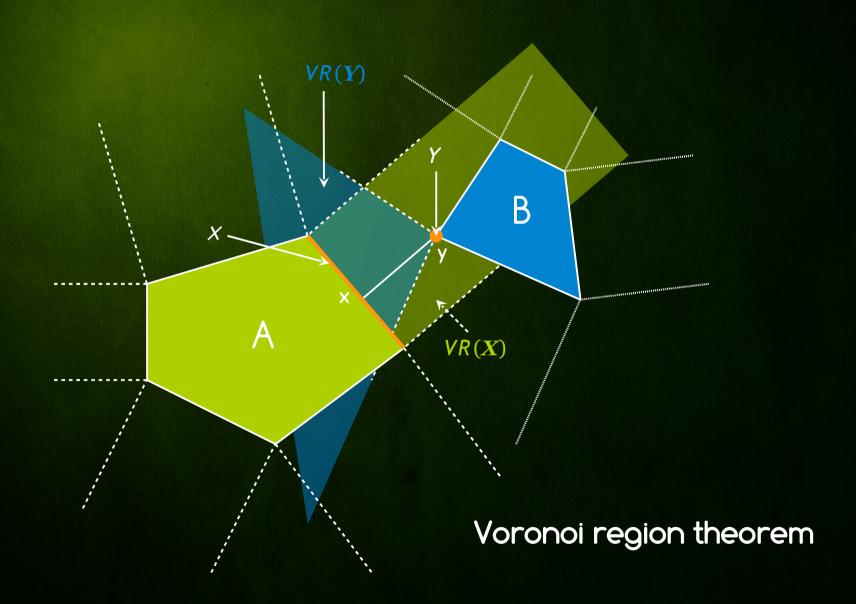
- → The inter-feature distance d(X, Y) between features X and Y is the minimum distance between any points $x \in X$ and $y \in Y$
- \rightarrow d(X,Y) = min { $|x-y| | x \in X \land y \in Y$ }



Voronoi Region Theorem

- *Let $x \in X$ and $y \in Y$ be the closest points between X and Y
- * Points x and y are the (globally) closest points between A and B iff $x \in VR(Y) \land y \in VR(X)$

Voronoi Region Theorem



V-Clip Algorithm

- * Key idea of the V-Clip algorithm is an efficient search for two closest features.
- * Obviously an exhaustive search is a very expensive solution
- *Fortunately the following Voronoi Region Theorem allows as to find the global minimum of the interfeature distance, by performing usually only a few iterations of a local search

V-Clip Algorithm

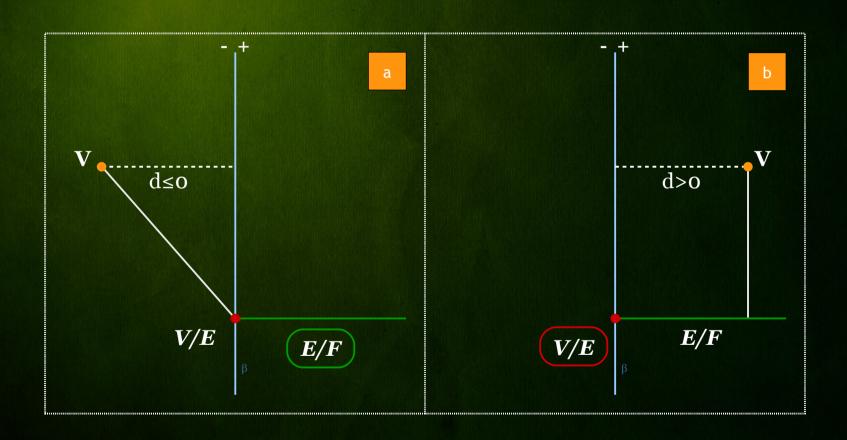
- Given two convex polytopes A, B and any two features X ∈ A, Y ∈ B
- * In each iteration V-Clip checks if they satisfy the Voronoi Region Theorem.
 - → If they don't, it changes X and Y to some (usually incident) features X' and Y', so that either the sum their dimensions or the inter-feature distance strictly decreases.
 - Assuming a finite number of features the algorithm can never cycle
 - → If we initialize X and Y with the closest features from the previous time-step and the simulation is coherent, then we probably need only a few iterations to find new closest features.

```
In: A pair of convex polytopes A, B and respective initial features X, Y
Out: A Separation vector w, or 0 if penetration occurred
function V-CLIP(A, B, X, Y): w
       while (true) do
             switch PairType(X, Y) do
 2:
                  case VV type:
                                                                                                                /* Vertex-Vertex */
 3:
                        if CLIPVERTEX(X, Y, \{ YE \mid E \in EDGES(Y) \}) then continue
 4:
                       if CLIPVERTEX(Y, X, \{XE \mid E \in EDGES(X)\}) then continue
 5:
                        return X - Y
 6:
 7:
                  end
                  case VE tupe:
                                                                                                                   /* Vertex-Edge */
 8:
                       if CLIPVERTEX(X, Y, \{ V_1^Y Y, V_2^Y Y, Y F_1^Y, Y F_2^Y \}) then continue
 9:
                       if CLIPEDGE(Y, X, \{XE \mid E \in EDGES(X)\}) then continue
10:
                       \mathbf{u} \leftarrow V_2^Y - V_1^Y
11:
                       \mathbf{return} \ X - \left(V_1^Y + \frac{\mathbf{u}^{\mathrm{T}}(X - V_1^Y)}{\mathbf{u}^{\mathrm{T}}\mathbf{u}}\mathbf{u}\right)
12:
13:
                  end
                  case VF type:
                                                                                                                    /* Vertex-Face */
14:
                       if CLIPVERTEX(X, Y, \{ EY, V_1^E E, V_2^E E \mid E \in EDGES(Y) \}) then continue
15:
                       if CLIPFACE(Y, X, A) then continue
16:
                       return X - \left(X + \frac{\mathbf{n}^{\mathrm{T}}(V_1^Y - X)}{\mathbf{n}^{\mathrm{T}}\mathbf{n}}\mathbf{n}\right)
17:
18:
                  end
                  case EE type:
                                                                                                                       /* Edge-Edge */
19:
                        if CLIPEDGE(X, Y, \{ V_1^Y Y, V_2^Y Y, Y F_1^Y, Y F_2^Y \}) then continue
20:
                       if \texttt{CLIPEDGE}(Y, X, \left\{ \begin{array}{l} V_1^X X, \ V_2^X X, \ XF_1^X, \ XF_2^{X'} \end{array} \right\}) then continue \left\{ \mathbf{u}^X, \ \mathbf{u}^Y \right\} \leftarrow \left\{ V_2^X - V_1^X, \ V_2^Y - V_1^Y \right\}
21:
22:
                        \{\mathbf{n}^X, \mathbf{n}^Y\} \leftarrow \{(\mathbf{u}^X \times \mathbf{u}^Y) \times \mathbf{u}^Y, (\mathbf{u}^Y \times \mathbf{u}^X) \times \mathbf{u}^X\}
23:
                       return \left(V_1^X + \frac{(\mathbf{n}^Y)^{\mathrm{T}}(V_1^Y - V_1^X)}{(\mathbf{n}^Y)^{\mathrm{T}}\mathbf{n}^X}\mathbf{u}^X\right) - \left(V_1^Y + \frac{(\mathbf{n}^X)^{\mathrm{T}}(V_1^X - V_1^Y)}{(\mathbf{n}^X)^{\mathrm{T}}\mathbf{u}^Y}\mathbf{u}^Y\right)
24:
25:
                  end
                  case EF tupe:
                                                                                                                       /* Edge-Face */
26:
                       if CLIPEDGE(X, Y, \{ EY, V_1^E E, V_2^E E \mid E \in EDGES(Y) \}) then continue
27:
                       \{d_1, d_2\} \leftarrow \{d_s(V_1^X, Y), d_s(V_2^X, Y)\}
28:
                        if (\operatorname{sgn}(d_1d_2) < 0) then Y \leftarrow \emptyset; continue
29:
                       if (|d_1| < |d_2|) then X \leftarrow V_1^X else X \leftarrow V_2^X
30:
                        continue
31:
                  end
32:
                  case EV, FV, FE type: SWAP(X,Y); SWAP(A,B); continue;
                                                                                                                 /* Swap Cases */
33:
             end
34:
             if (Y = \emptyset) then return \emptyset
35:
36:
       end
end
```

Vertex Clipping

- Given a vertex V from one object, some "old" feature N from another object and a set of feature pairs S_n
- * The vertex clipping simply marks X (Y) if the vertex V lies above (below) the VP(X,Y) for each feature pair XY \in S_N
 - → First it clears all features among SN (ClearAll(S_N))
 - → Next it tests the side (w.r.t. Voronoi plane) of V and mark "further" features.
 - → Finally it updates N with some unmarked feature (UpdateClear (N, SN)) and returns true if N was changed.

Vertex Clipping Cases



ClipVertex and UpdateClear

```
In: A vertex V, a feature N to be updated and a set of clipping feature pairs S_N
Out: Test if the feature N was updated (true/false)
function CLIPVERTEX(V, N, S_N): bool
      CLEARALL(S_N)
 1:
      for each XY in S_N do
 2:
          Test \leftarrow \operatorname{sgn}(\operatorname{d}_{\mathbf{s}}(V, \mathcal{VP}(X, Y)))
 3:
          if (Test > 0) then Mark(X) else Mark(Y)
 4:
      end
 5:
      return UPDATECLEAR(N, \mathcal{S}_N)
 6:
end
In: A feature N to be updated and a set of clipping feature pairs S_N
Out: Test if the feature N was updated (true/false)
function UPDATECLEAR(N, S_N): bool
     M \leftarrow N;
1:
                                                       /* store old feature */
     foreach XY in S_N do
2:
        if (X \text{ is "} clear") \text{ then } N \leftarrow X; \text{ break;}
3:
                                                       /* update old to closest feature */
        if (Y \text{ is "} clear") \text{ then } N \leftarrow Y; \text{ break;}
                                                       /* update old to closest feature */
4:
     end
5:
     return N := M;
                                                       /* true if feature changed */
```

end

Edge Clipping

- * Take an edge E, the "old" feature N, a set of respective feature pairs S_N and perform a sequence of local tests to properly mark "further" features
- *Let d_1 , d_2 represent signed distances of the endpoint vertices V_1^E , V_2^E to the Voronoi plane β = VP(X,Y) of a particular feature pair XY \in S_N
- * If both vertices lie on the same side of the clipping plane ($sgn(d_1d_2) > 0$), we simply mark the feature of the opposite side as in vertex clipping

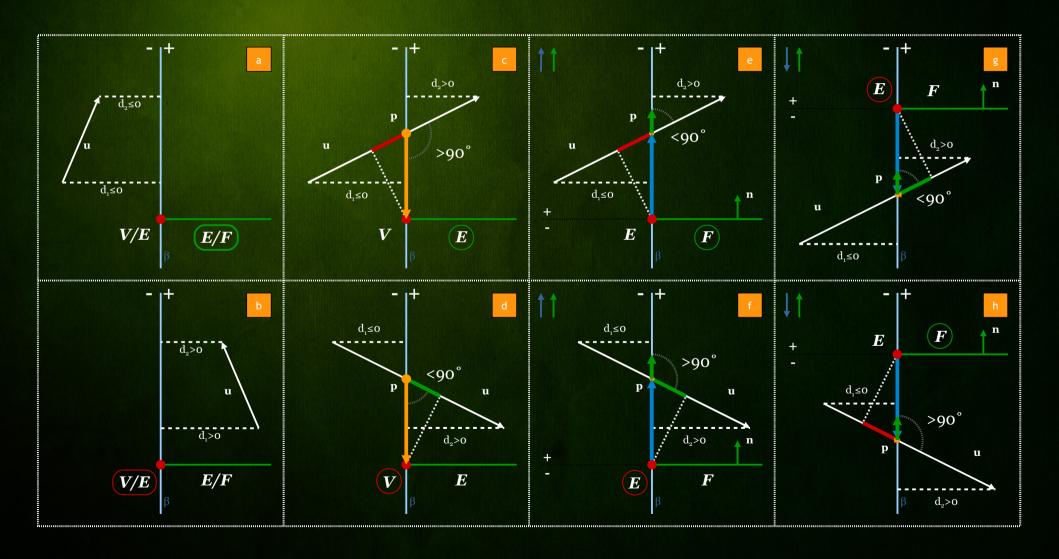
Edge Clipping

- * If vertices lie on different sides ($sgn(d_1d_2) < 0$), edge E intersects the clipping plane in some point $\rho = (1 \lambda)V_1^E + \lambda V_2^E$, where $\lambda = d_2/(d_1-d_2)$ and we must consider two sub-cases depending on the type of the feature pair
- *Let vector $u = sgn(d_2)(V_2^E V_1^E)$ represent the edge E pointing out of the negative half-space to the positive half-space of β
- * If XY is a "VE" pair, the local test depends on the sign of the $(X \rho)$ projection onto the edge vector u, i.e. $+sgn(u^T(X \rho))$

Edge Clipping

- * If XY is a "EF" pair, there are another two subcases.
- * If p lies above the face Y, the local test depends on the angle between edge vector u and the face normal vector n
- * If p lies below the face Y we use the similar local test, but mark opposite features
- *Therefore the final local test (handling both subcases) can be written as: sgn(n^Tu)sgn(d_s(p,Y))

Edge Clipping Cases



ClipEdge Algorithm

```
In: An edge E, a feature N to be updated and a set of clipping feature pairs S_N
Out: Test if the feature N was updated (true/false)
function CLIPEDGE(E, N, S_N): bool
         CLEARALL(S_N)
 1:
         for each XY in S_N do
 2:
              \beta \leftarrow \mathcal{VP}(X,Y)
 3:
              \{d_1, d_2\} \leftarrow \{ d_s(V_1^E, \beta), d_s(V_2^E, \beta) \}
                                                                                                           /* signed distances to \beta
 4:
              \{\mathbf{p}, \mathbf{u}\} \leftarrow \{ E(d_2/(d_1 - d_2)), \operatorname{sgn}(d_2)(V_2^E - V_1^E) \}
  5:
              if (\operatorname{sgn}(d_1d_2) > 0) then Test \leftarrow \operatorname{sgn}(d_1)
  6:
              if (\operatorname{sgn}(d_1d_2) < 0 \land XY \text{ is "}VE") then \operatorname{Test} \leftarrow +\operatorname{sgn}(\mathbf{u}^{\mathrm{T}}(X - \mathbf{p}))
 7:
              if (\operatorname{sgn}(d_1d_2) < 0 \land XY \text{ is "}EF") then \operatorname{\mathsf{Test}} \leftarrow -\operatorname{\mathsf{sgn}}(\mathbf{n}^{\mathrm{T}}\mathbf{u})\operatorname{\mathsf{sgn}}(\operatorname{d}_{\mathbf{s}}(\mathbf{p},Y))
 8:
              if (Test > 0) then Mark(X) else Mark(Y)
 9:
10:
         end
         return UPDATECLEAR(N, \mathcal{S}_N)
11:
end
```



Signed Distance Map

- * Signed distance map: $SDM_N(V)$ is N*N*N regular grid, where each unit cell with a center point p stores the signed distance to the closest point on the surface of some volume V.
- * This signed distance is a combination of a sign function $sgn_{v}(\rho)$ and the unsigned distance function $d(\rho, V)$ w.r.t. V.
 - → $SDM_N(V) = \{ sgn_V(\rho)d(\rho,V) \mid \rho = (i + 0.5, j + 0.5, k + 0.5) \land 1 \le i, j, k \le N \}$

Signed Distance Maps

- * Signed distance maps (SDM) become recently a popular technique for approximate collision detection and distance computation.
- * Pros: Efficient overlap test, fast contact generation and penetration depth computation for arbitrary shaped, non-convex objects with complex and highly tessellated geometry
- * Suitable even for real-time applications as games
- * Cons: Huge amount of memory necessary for massive scenarios and a large number of redundant (unnecessary) contacts generated during the collision detection

Distance Map Construction

* Brute force construction

- For each grid cell we need to compute the distance of its center to each surface triangle and store the shortest distance
- Assuming N is the grid size and M is the number of triangles, we have to call the primitive point-to-triangle distance function N*N*N*M times

* Other Efficient Methods

- Lower-Upper Bound Tree (LUB-Tree)
- Characteristic/Scan Conversion (CSC)
- Chamfer and Vector Distance Transform (CDT, VDT)
- Fast Marching Method (FMM)

Proximity Queries with SDM

- * Performing proximity queries using SDM involves simple point location tests.
- * The key idea is to sample several points on the surface and store it together with the SDM.
- * During the collision detection sample points of one object are transformed into the local space of the other object and are "looked-up" in the SDM of the other object and vice versa.
- * Surface points located inside other object (lie under the zero level (SDM_A(p_B) ≤ 0)) are used to create necessary contact information (contact point, contact normal, penetration depth, etc.)

