Algorithmic aspects of alternating sum of volumes. Part 2: Nonconvergence and its remedy

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The paper is the second part of a 2-part paper. The first part focused on the issues of data structure and fast difference operation. The second studies the nonconvergence of the alternating sum of volumes (ASV) process. An ASV is a series of convex components joined by alternating union and difference operations. It is desirable that an ASV series be finite. However, such is not always the case – the ASV algorithm can be nonconvergent. The paper investigates the causes of this nonconvergence, and finds and proves the conditions responsible for it. Linear time algorithms are then developed for detection.

leature extraction, representation conversion, convex hull, alternating sum, difference operation, nonconvergence, supportability

An alternating sum of volumes (ASV) series is convergent if a deficiency Ω_n is the null set; otherwise, it is said to be nonconvergent (For the computation of efficiency, the detection of a null deficiency Ω_n can be replaced by the determination of the convexity of Ω_{n-1}) Figure 1 illustrates a nonconvergent ASV series. The series of deficiencies $\Omega_1, \Omega_2, \ldots$, as derived from the convex hull (CH) and difference (-) operations, never converges to the null set, resulting in an infinite alternating series. {CH(Ω_1) - CH(Ω_1) + CH(Ω_2) - \cdots - CH(Ω_{2i-1}) + CH(Ω_{2i}) - \cdot }

As implied in Figure 1, the nonconvergence of an ASV series is determined by the nonconvergence of a deficiency in its expansion. It is known¹ that an ASV series is nonconvergent when the convex hull of a deficiency Ω_{i} is identified with the convex hull of the deficiency of Ω_{i+1} . For the example shown in Figure 1, the convex hull $CH(\Omega_1)$ is equal to the convex hull $CH(\Omega_2)$. As a result of the identification $CH(\Omega_i) = CH(\Omega_{i+1})$, the following relationship between the deficiencies holds $\Omega_1 = \Omega_{j+2}$ ($i \leq j$).

Formally, a deficiency $\hat{\Omega}$, is said to be nonconvergent

if the convex hull of its deficiency $CH(\mathbf{\Omega}_i) - \mathbf{\Omega}_i$ is equal to $CH(\mathbf{\Omega}_i)$, and convergent otherwise. It is desirable to be able to characterize the nonconvergence of a deficiency Ω , directly, rather than invoke the comparison between $CH(\mathbf{\Omega}_i)$ and $CH(\mathbf{\Omega}_i) - \mathbf{\Omega}_i$. This pursuit is justified in two respects. First, four convex-hull operations and two set-difference operations must be performed to obtain the datum $CH(\Omega_i)$, $CH(\Omega_i) - \Omega_{ii}$ and $CH(CH(\mathbf{\Omega}_i) - \mathbf{\Omega}_i)$ for the comparison. Setdifference operation on a polyhedron with m vertices is known to take at least $O(m^2)$ time prior to the $O(m\log m)$ result given in Part 1 of this paper². Second, even if the fast $O(m \log m)$ difference operation is involved, detecting the presence of a null set, as the result of the difference, can be numerically unstable A fast nonconvergence detection algorithm for a



Figure 1 Illustration of ASV nonconvergence [CH convex-hull operation, – difference operation]

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pseudopolyhedron, where the set-difference and comparison operations are not carried out, is a new result in this Part 2 of the paper

Suppose that a fast nonconvergence detection algorithm for a deficiency is available. One way in which to detect the nonconvergence of an ASV series is to test for the nonconvergence of every deficiency as it is being computed. The time required by such a detection scheme is heavily dependent on the depth nof the first nonconvergent deficiency Ω_n – the larger the number *n*, the more time it will take. Alternatively, it may be asked whether the nonconvergence of a series can be detected without invoking the ASV process itself – not only because the deficiencies thus produced are nonproductive if the ASV series does not converge, but also because a separate scheme may speed up the detection time. From the theoretical point of view, such a study addresses some interesting problems, such as that of finding the minimum number of faces in a nonconvergent deficiency

These two closely related issues, fast detection of the nonconvergence of a deficiency and that of an ASV series, are investigated in this paper. In the next section, the concepts of *strong-hull* and *weak-hull vertices* are introduced. The characterization of these two types of vertices leads to an O(nlogn) algorithm for detecting the nonconvergence of a deficiency, where *n* is the number of vertices in the deficiency. In the third section of the paper, a sufficient condition for the nonconvergence of an ASV series that requires only linear time to detect is given

CHARACTERIZATION OF NONCONVERGENT DEFICIENCIES

The following problem is solved in this section given a pseudopolyhedron Ω_{μ} under what condition will the equation $CH(\mathbf{\Omega}_i) = CH(CH(\mathbf{\Omega}_i) - \mathbf{\Omega}_i)$ hold, and how fast can such a condition be detected? The symbols CH and - represent the convex hull and regularized difference operations, respectively (Note that every deficiency in an ASV series must be a pseudopolyhedron, as shown in Part 1 of this paper Hereafter, the two terms 'pseudopolyhedron' and 'deficiency' will be used interchangeably) Before the condition for nonconvergence is characterized, it is useful to summarize the relationships between the boundary and interior points of a pseudopolyhedron Ω_{ij} its convex hull $CH(\Omega_i)$, and its deficiency $CH(\Omega_i) - \Omega_i$. The first relationship, given in Part 1 of this paper, is recited below

Lemma 1 The deficiency of a pseudopolyhedron Ω_i is also a pseudopolyhedron, whose interior $I(CH(\Omega_i) - \Omega_i)$ is the set difference $\{I(CH(\Omega_i)) - I(\Omega_i)\}$, and the boundary $B(CH(\Omega_i) - \Omega_i)$ is a subset of $\{B(CH(\Omega_i)) - B(\Omega_i)\}$ that forms the closure of $\{I(CH(\Omega_i)) - I(\Omega_i)\}$

A pseudopolyhedron is completely described by its faces, and a face is determined by its edges, which are themselves defined by their end points, called vertices As the set of the vertices of the convex hull of a set of points must be a subset of that point set, by Lemma 1,

the set of the vertices of the deficiency of Ω_i is a subset of the vertices of Ω_i . In other words, the difference operation in the ASV expansion can be viewed as a vertex-elimination process after each difference operation, the deficiency Ω_i possesses fewer vertices than does the deficiency Ω_{i-1} , this process continues until a convex pseudopolyhedron Ω_n is reached whose deficiency Ω_{n+1} is the null set

If the vertices in the deficiencies cannot be eliminated through the difference operation, the ASV series does not converge A vertex of a pseudopolyhedron Ω_i is eliminatable if it does not exist in its deficiency $CH(\Omega_i) - \Omega_i$, otherwise, it is noneliminatable. A formal definition of the nonconvergence of the pseudopolyhedron is then in order

Definition 1 A pseudopolyhedron Ω , is nonconvergent if all of its vertices are noneliminatable, otherwise, it is convergent

To characterize the eliminatability of vertices in Ω_i , the vertices are categorized into two groups, *hull vertices* and *internal vertices*. The hull vertices are those that are on the boundary of $CH(\Omega_i)$, whereas those vertices of Ω_i that are not on the boundary of $CH(\Omega_i)$ are internal. Each of the internal vertices has a 3D neighborhood that is strictly inside $CH(\Omega_i)$ Further, this neighborhood contains a subset of $\{I(CH(\Omega_i)) - I(\Omega_i)\}$, as an internal vertex is also a boundary point of Ω_i . Therefore, by Lemma 1, all the internal vertices are noneliminatable. To study the eliminatability of the hull vertices, they are further separated into *weak* and *strong* hull vertices.

Definition 2 In E³, the 3D Euclidean space, a hull vertex of Ω , is weak if it has a 3D neighborhood that contains points in $\{\Omega_i \cup \{E^3 - CH(\Omega_i)\}\}$ only, otherwise, it is called a strong hull vertex

As shown in Figure 2, after a difference operation, strong hull and internal vertices remain, whereas all the weak hull vertices are eliminated. Let those faces (edges) of a pseudopolyhedron Ω , be called *hull faces* (*hull edges*) if they are completely on the boundary surface of $CH(\Omega_i)$, and *internal faces* (*internal edges*) otherwise Referring to Figure 2, it can be inferred that a hull vertex is weak if and only if all of its incident faces are hull faces of Ω_i (Note, however, that this condition does not hold for incident edges. That is, a hull vertex with incident hull edges only is not necessarily weak, as shown in Figure 3, where the strong hull vertex *v* has no incident internal edges.) The contribution of strong hull vertices to the nonconvergence is demonstrated by the following lemma



Figure 2. Weak hull vertices, strong hull vertices and internal vertices

[O weak hull vertices, • strong hull vertices, × internal vertices]



Figure 3 Strong hull vertex with no incident internal edges

Lemma 2. A pseudopolyhedron Ω_i is nonconvergent if and only if all of its hull vertices are strong.

Proof. First, it is noted that the hull and internal vertices partition the entire vertex set of Ω_i , owing to their mutual exclusivity By Definition 2, a weak hull vertex has an open 3D neighborhood within which Ω_i is equal to $CH(\Omega_i)$, and thus there is no subset of $\{I(CH(\Omega_i)) - I(\Omega_i)\}$ in that neighborhood Hence, by Lemma 1, all the weak hull vertices are eliminatable Conversely, as every 3D neighborhood of a strong hull vertex contains a subset of $\{I(CH(\Omega_i)) - I(\Omega_i)\}$, these are preserved on the deficiency of Ω_i , i.e. they are noneliminatable By Definition 1 and the fact that all the internal vertices are noneliminatable, the proof is complete QED

Lemma 2 implies that the detection of the nonconvergence of a pseudopolyhedron Ω , is equivalent to distinguishing its strong hull vertices from the weak ones Such a process requires two steps classify the hull and internal faces of Ω_i , and then check if Ω_i has a vertex that has incident hull faces only. Whether a face is internal can be identified by checking one of its interior points. (Such a point must not be on an edge of the face, as an internal face may have hull edges only, e.g. face f in Figure 3.) Ω_i is then nonconvergent if and only if no weak hull vertex exists.

The algorithm given below follows the two steps just described It is assumed that a procedure HULL(N, V, V_{tag}) is in hand that takes a list V of N points as input, and outputs a property array V_{tag} such that, if $V_{tag}(i)$ is 'true', point i in V is a hull vertex of CH(V), and if it is 'false', an internal vertex.

DETECT algorithm

Algorithm DETECT (Ω_i) /* Detect the nonconvergence of a pseudopolyhedron Ω_i The vertex list V and face list F of Ω_i have n_v vertices and n_i faces, respectively */		
begin (1)	for $k = 1$ to n_f do $V(n_y + k) \leftarrow an$ interior point of face k in F end do	
(2)	call HULL($n_v + n_l, V, V_{tag}$)	
(3)	set array VP(1 n_v) to 'true'	
(4)	for $k = 1$ to n_l do for every vertex v of face k in F do $VP(v) \leftarrow VP(v) \cap V(n_v + k)$ end do end do	
(5)	<pre>for k = 1 to n, do if VP(k) = 'true' then return ('convergence) end if end do</pre>	
(6) end D	return ('nonconvergence') IETECT	

In the DETECT algorithm, the $n_{\rm f}$ interior points of the faces of $\mathbf{\Omega}$, are first appended to the vertex array V of $\mathbf{\Omega}_{\mu}$. As each interior point of a face can be obtained in constant time by considering any two adjacent edges of that face, Step 1 takes $O(n_f)$ time. The convex-hull procedure HULL is called at Step 2, which requires only $O((n_v + n_f)\log(n_v + n_f))$ time³ At Step 3, a property array $VP(1:n_v)$ is preset to 'true'. At Step 4, the following is carried out if a face k is internal, i.e. its interior point tag $V_{\text{tag}}(n_v + k)$ is 'false', the corresponding entries in VP for all the vertices of face k are reset to 'false' Such a process obviously takes O(D) time, where $D = \Sigma d_i$ $(i = 1, 2, ..., n_v)$, and d_i is the degree of vertex *i* it is shown in the Appendix of Part 1 of this paper that D is $O(n_f)$. Finally, at Step 5, the array VP is scanned, and Ω_{i} is identified as convergent if some entry in VP is 'true', and as nonconvergent otherwise. The time complexity of the DETECT algorithm is summarized by the following theorem.

Theorem 1 The detection of the nonconvergence of a pseudopolyhedron Ω , with *n* vertices can be done in $O(n\log n)$ time.

Compared with the simple comparison method¹, $CH(\Omega_i) = CH(CH(\Omega_i) - \Omega_i)$, the new detection algorithm DETECT avoids both the time-consuming difference operation and the identification of a null set that could be numerically unstable Two convex-hull operations are also saved

It may be noted that the detection algorithm DETECT disregards the disconnectedness of a set The pseudopolyhedron Ω_{i} in Figure 4(a) is nonconvergent, by Lemma 2 The deficiency Ω_{i+1} , however, consists of two separate pseudopolyhedra \mathbf{P}_1 and \mathbf{P}_2 . Although Ω_{i+1} is nonconvergent as a single set, it is convergent if represented as $ASV(\Omega_{i+1}) = ASV(\mathbf{P}_1 + \mathbf{P}_2) = ASV(\mathbf{P}_1) + ASV(\mathbf{P}_2)$, because \mathbf{P}_1 and \mathbf{P}_2 are both convergent it results in a convergent ASV tree $ASV(\Omega_i) = \mathbf{H} - \Omega_{i+1} = \mathbf{H} - ASV(\Omega_{i+1}) = \mathbf{H} - (ASV(\mathbf{P}_1) + ASV(\mathbf{P}_2))$, which branches at the deficiency Ω_{i+1} in some other cases, a pseudo-polyhedron, though connected, might be separated uniquely at some edges such that the separated subsets



Figure 4 Convergence by set separation



Figure 5 Well connected set versus ill connected set, (a) well connected set, (b) ill connected set

are all convergent. For example, the pseudopolyhedron Ω , in Figure 4(b) is nonconvergent, by Lemma 2. It is, however, convergent if expressed as $\Omega_{i} = H - \Omega_{i+1} = H - (P_1 + P_2 + P_3 + P_4)$, as P_1 , P_2 , P_3 and P_4 are all convergent.

Both examples of set separation on Ω_{i+1} shown in Figure 4 have a crucial property the boundary of the separated pseudopolyhedron remains unchanged Unlike in polyhedral decomposition⁴, such a property guarantees that the boundary after the separation has the same sets of vertices, edges and faces as before, with only the adjacency and incidence relationships between them being altered. Further, it is next shown that such a separation is unique, thus justifying the existence of a deterministic algorithm. To define set separation rigorously, the concept of well connectedness is needed

Definition 3 Two points p and q of a pseudopolyhedron Ω , are said to be well connected in Ω , if there exists a curve c between p and q such that all the points in c, except for possibly p and q, are in $I(\Omega_i) \ \Omega_i$ is a well connected set if all of its points are well connected in it, otherwise it is an *ill connected* set (see Figure 5)

The Ω_{i+1} in Figure 4(a), and both Ω_i and Ω_{i+1} in Figure 4(b), are ill connected pseudopolyhedra A well connected pseudopolyhedron is also called a *robust* set, meaning that its interior is all connected A subset ζ of a pseudopolyhedron Ω_i is a maximally well connected set (MWCS) of Ω_i if ζ is a well connected set, and any addition of non- ζ points of Ω_i to ζ will constitute an ill connected set As an example, only P_1 , P_2 , P_3 and P_4 are the MWCSs of the pseudopolyhedron Ω_{i+1} in Figure 4(b)

It is desirable that an ASV series be expanded as much as possible so that more features can be extracted Once a nonconvergent and ill connected deficiency is encountered, it should be separated into the MWCSs, and the ASV process can then be performed on each of these This leads to the notion of *strong* and *weak* nonconvergence.

Definition 4 A nonconvergent pseudopolyhedron Ω , is strongly nonconvergent if both itself and its deficiency are robust. Otherwise, Ω , is weakly nonconvergent

As examples, the deficiency Ω_1 in Figure 1 is strongly nonconvergent, as both it and its deficiency Ω_2 are robust, whereas- each Ω_1 , in Figure 4 is weakly nonconvergent, because either it, or its deficiency Ω_{t+1} , is ill connected.

The detection of the strength of nonconvergence of a pseudopolyhedron Ω_i , of *n* faces requires three steps the identification of its nonconvergence, the computation for the deficiency of Ω_i , and the classification of the well connectedness of Ω_i and/or

Its deficiency. The first step can be done, by Theorem 1, in O(nlogn) time. The difference operation also requires O(nlogn) time, as shown in Part 1 of this paper. Thus, if the classification of the robustness of Ω_i can be done in O(nlogn) time, then so can the detection of the strong nonconvergence. As Ω_i is robust if and only if its MWCS separation contains only one MWCS, i.e. itself, the goal becomes the finding of an O(nlogn) time MWCS separation algorithm

In implementing such an algorithm, it is noted that, by the definition of a pseudopolyhedron, the well connectedness of its boundary-point set will ensure that it is a well connected set. This fact ensures that the MWCSs of a pseudopolyhedron can be detected by checking only the well connectedness of its faces.

Let two faces of a pseudopolyhedron be *well adjacent* to each other if they share a common edge and are well connected to each other. It can easily be shown that two faces A and B of a pseudopolyhedron are well connected if and only if either they are well adjacent, or there exists a number of faces f_1, f_2, \ldots, f_k such that A is well adjacent to f_1, f_1 is well adjacent to f_2, \ldots, f_k such that A is well adjacent to B For example, in Figure 6, faces A and B are not well connected, because the curve c connecting points p and q passes through edge e, which does not belong to the interior of that pseudopolyhedron.

To characterize face well adjacency, let f_1, f_2, \dots, f_m be the faces incident to a common edge, ordered by their spatial angles (In Figure 7, f_1, f_2, \dots, f_m are the intersections between the faces sharing a common



Figure 6 Ill connectedness of faces of a pseudopolyhedron



Figure 7 Well adjacency of taces

edge and a plane orthogonal to that edge) Apparently, the well adjacent face of face f_i (i = 1, 2, ..., m) is either f_{i-1} or f_{i+1} (mod m), depending on the direction of the outward normal of f_i . Such a pairing process can be done in O(m) time. The following recursive procedure MWCS_FACES finds the faces of a maximally well connected set of a pseudopolyhedron Ω_i . The input is the pseudopolyhedron representation {V, E, F,NORM, E_f } of Ω_i , and the index of a face of Ω_i . The output is the indices of those faces of Ω_i that form the boundary of an MWCS of Ω_i .

MWCS_FACES procedure

Procedure MWCS_FACES (I, Ω _i) /* Find those faces of a maximum well connected set of pseudopolyh (is the index of a face that is required to be on the MWCS	edron $\mathbf{\Omega}$,
• /	
begin	
(1) output (
(2) $e_1 e_2$, $e_k \leftarrow edges of face l$	
(3) for $i = 1, k$ do	
(3.1) I' \leftarrow the index of the face well adjacent to I at edge e,	
(3.2) If (I' has not been output) then call MWCS_FACES(I', Ω_i)	
end do {Step 3}	
end MWCS_FACES	

, e_m of $oldsymbol{\Omega}_{,}$ Suppose that a total of m edges e_1 , e_2 , are found in an MWCS, denoted by P, through MWCS_FACES Let k_i and k'_i be the face-adjacency indices of Ω_i , and P at edge e_i, respectively (i = 1, 2, ..., m) Step 1 takes constant time, and thus the overall time spent at Step 1 when MWCS_FACES terminates is O(n), where n is the number of faces on P. As each face of P is processed only once, the overall , m)) As time required by Step 2 is $O(\Sigma(k'_i)) = 1, 2,$ for the loop at Step 3, note that the indices of the faces of Ω_i adjacent to edge e_i are stored in the order of their spatial angles in an entry of the E_f list of Ω_i Therefore, only O(logk,) time is needed to locate the position of f in that entry, and, hence, the index f' of the face well adjacent to face f at an edge e_i . As a result, the overall time taken by Step 3.1 is $O(\Sigma(k'_l \log k_l))$ (m)) The total time cost of MWCS_FACES (i = 1, 2, ...)is therefore $O(n + (\Sigma(k'_i \log k_i) (i = 1, 2, ..., m)))$.

Before the complete algorithm for carrying out the MWCS separation is presented, it is necessary to clarify that, given the indices of *n* faces that form the boundary of an MWCS of Ω_{μ} only O(n) time is needed to construct the pseudopolyhedron representation $\langle V, E, F, NORM, E_f \rangle$ of that MWCS, say P_i . To see this, note that all the V, E, F, NORM and E_f lists of P, are readily available in the $\langle V, E, F, NORM, E_f \rangle$ of Ω , The only work needed besides the retrieval is the reindexation of the vertices, edges and faces of P, once they are retrieved from Ω_i . For example, if only vertices $\{v_{3}, v_{4}, v_{7}, v_{9}, v_{15}\}$ are on P_{1} , and there is an edge on P, whose entry in the E list of $\mathbf{\Omega}_{i}$ is $\langle 9, 4 \rangle$, then this edge will become $\langle 4, 2 \rangle$ in the *E* list of *P*₁, because vertices v_9 and v_4 now sit at the fourth and second positions of the V list of P, Analogously, if edges $\{e_2, e_7, e_{10}, e_{13}\}$ are on P_1 , and P_1 has a face stored in the F list of Ω , as $\langle 10, 7, 13 \rangle$, then this face will become $\langle 3, 2, 4 \rangle$, owing to the reindexation of $\{e_2, e_7, e_{10}, e_{13}\}$. Clearly, this reindexation process can be done in O(n) time through simple index mapping Let MWCS_OUTPUT be such a process, that takes as input a pseudopolyhedron Ω_i , and a list L of indices of the faces of Ω_i , and outputs the pseudopolyhedron representation of an MWCS of Ω_i , whose faces are those of Ω_i , with indices in L. Using both procedures MWCS_FACES and MWCS_OUTPUT, the algorithm given next performs the MWCS separation

MWCS_SEPARATION algorithm

Algorithm MWCS_SEPARATION ($\mathbf{\Omega}_{i}$)			
Compute the MWCSs of a pseudopolyhedron $\mathbf{\Omega}$, and output them	•/		
begin			
(1) unmark all the faces in the F list of Ω_{i}			
(2) while (there is a face I in F which is not marked) do			
(2.1) $L \leftarrow MWCS_FACES(f, \Omega_i)$			
(2.2) call MWCS_OUTPUT(L, Ω_i)			
(2.3) mark all the faces with indices in L			
end do {Step 2}			
end MWCS_SEPARATION			

Lemma 3. The MWCS separation of a pseudopolyhedron Ω_i , with n_f faces can be done in $O(n_f \log n_f)$ time and $O(n_f)$ space

Proof. In the MWCS_SEPARATION algorithm, Step 1 takes $O(n_f)$ time. For the while loop at Step 2, as each face can only be in one MWCS, the overall time cost of Step 2.2 and Step 2.3 is clearly $O(n_f)$ The time taken by each execution of the procedure MWCS_FACES is in the form of $O(n + (\Sigma(k'_{i}\log k_{i}) \ (i = 1, 2, ..., m)))$, where *n* and *m* are the numbers of the faces and edges on that particular MWCS, and k_i and k'_i are the numbers of the faces of Ω , and that MWCS adjacent to an edge of the MWCS, respectively. For the same reason that a face of $\mathbf{\Omega}$, can only be in one of its MWCSs, the sum of $\Sigma k'_i$ (i = 1, 2, ..., m) over all the edges of Ω_i is $O(\Sigma k_i)$ $(i = 1, 2, ..., n_e))$, where n_e is the total number of edges of Ω , Therefore, after the termination of the MWCS_SEPARATION algorithm, the overall time taken by Step 2.1 is $O(n_f + (\Sigma k_i) \log n_f)$, i.e. $O(n_f \log n_f)$, as Σk_i (i = 1, 2, ..., n_e) is $O(n_f)$. QED

With Theorem 1 and Lemma 3, the following is in order.

Theorem 2: Whether a pseudopolyhedron Ω_i , is strongly nonconvergent or not can be detected in $O(n\log n)$ time, where n is the number of faces of Ω_i

It is worth noting that, in the ASV process, the MWCS_SEPARATION algorithm not only detects the strong nonconvergence of a deficiency Ω_i , but also constructs the MWCSs of the deficiency Ω_{i+1} . The pseudopolyhedron representation of the MWCSs can then be used for the subsequent convex-hull and difference operations, along the corresponding branches after Ω_{i+1} .

FAST DETECTION FOR ASV NONCONVERGENCE

An ASV series is nonconvergent if it has a nonconvergent deficiency Ω_n . A way to detect the nonconvergence of an ASV series is to check the nonconvergence of every deficiency in the series. The time required by such a detection sets an upper bound

Theorem 3 It needs at most $O(n^2 \log n)$ time to decide whether the ASV series of a pseudopolyhedron Ω is



Figure 8. Regularized set intersection

convergent or not, where n is the number of vertices of ${f \Omega}$

Proof Recall that the difference operation is a vertex-elimination process. That is, a nonconvergent deficiency Ω_k in ASV(Ω) always has fewer vertices than Ω_{k-1} . In the worst case, suppose that only one vertex is eliminated after each difference operation. To obtain the deficiency Ω_k through the ASV process, k convex-hull and difference operations are needed, resulting in an overall time requirement of $\Sigma O(i\log i)$ (i = n, n - 1, ..., n - k). Therefore, at most $\Sigma O(i\log i)$ (i = n, n - 1, ..., n - k). Therefore, at most $\Sigma O(i\log i)$ (i = n, n - 1, ..., n - k) converges or not. QED

In an attempt to improve this upper bound, the local cause of the ASV nonconvergence of a pseudopolyhedron Ω is sought Such a study results in a sufficient condition for the ASV nonconvergence, which eventually leads to a linear detection algorithm. In the search for this local cause, it is useful to invoke the mechanics of *regularized intersection*⁵

Definition 5 The regularized intersection of two pseudomanifolds **A** and **B**, denoted by $\mathbf{A} * \mathbf{B}$, is a pseudomanifold whose interior is $\mathbf{I}(\mathbf{A}) \cap \mathbf{I}(\mathbf{B})$

Figure 8 gives two examples of regularized intersection. As shown in Figure 8(a), the regularized intersection $\mathbf{A} * \mathbf{B}$ is the null set \emptyset , even though the ordinary set intersection $\mathbf{A} \cap \mathbf{B}$ yields two faces. The result of the regularized intersection in Figure 8(b) is a nonconvergent pseudopolyhedron. A tantalizing finding is revealed by the example in Figure 8(b) if there exists a (nonempty) subset (prior to the regularized intersection) that is nonconvergent, then the ASV series to be expanded is nonconvergent. Such an observation is not a coincidence, the basis of this is shown by the following Lemma 4

Lemma 4. Let ζ be a subset of the vertices of a pseudopolyhedron Ω . If the regularized intersection between Ω and CH(ζ) is a nonconvergent pseudopolyhedron, the ASV series of Ω is nonconvergent.

Proof Assume that $\mathbf{\Omega} \cdot CH(\zeta)$ is a nonconvergent pseudopolyhedron It is claimed that all the vertices in ζ are noneliminatable. Suppose that there is a deficiency $\mathbf{\Omega}$, in ASV($\mathbf{\Omega}$) whose vertex set is a superset of ζ , such that some vertex v in ζ is lost on the deficiency $\mathbf{\Omega}_{r+1}$. By Lemma 2, this means that all the incident faces of v are the hull faces of $\mathbf{\Omega}_r$. As $CH(\zeta)$ is a subset of

Lemma 4 provides a sufficient condition for the nonconvergence of an ASV series, without the ASV process itself being invoked. A direct implementation of such an algorithm is, however, infeasible, as there are O(n!) subsets. To reduce this high complexity, the characterization of local subsets of vertices, i.e. those that are adjacent to a common vertex, is investigated.

Let two vertices of a pseudopolyhedron be said to be *adjacent* to each other if they are the two end points of an edge

Definition 6 A vertex v of a pseudopolyhedron Ω is supportable if there exists a plane containing v such that the point set ξ_v lies on one side of that plane (where ξ_v consists of those vertices that are adjacent to v), otherwise, v is a nonsupportable vertex

As an example, all the vertices except for v of the pseudopolyhedron in Figure 9(a) are supportable. Also, as shown in Figure 9(b), a vertex v is nonsupportable if and only if it is strictly inside the convex hull of the vertices adjacent to it

Lemma 5 If a pseudopolyhedron Ω has a non-supportable vertex, then the ASV series of Ω is nonconvergent

Proof Let v be a nonsupportable vertex of Ω , and ξ_v be the point set consisting of those vertices that are adjacent to v The lemma is proven by showing that $\mathbf{\Omega} * CH(\xi_v)$ is a nonconvergent pseudopolyhedron As v is internal to CH(ξ_v), all its incident faces have portions that are internal to $CH(\xi_v)$ Then, in each of these faces, there is a point that has an open 3D neighborhood that contains a subset of $I(\Omega)$ that is strictly inside CH(ξ_v). By the definition of the regularized intersection, this neighborhood is preserved on $\mathbf{\Omega} * CH(\xi_v)$ In other words, $\mathbf{\Omega} * CH(\xi_v)$ must be a pseudopolyhedron, as its interior is not empty. Now, consider a hull vertex p of $\mathbf{\Omega} * CH(\xi_v)$, as illustrated in Figure 10(a) If p belongs to $\xi_{\rm v}$, as none of the incident faces of v can be a hull face of CH(ξ_v), p can only be a strong hull vertex of $\mathbf{\Omega} * CH(\xi_v)$ If p does not belong to ξ_{v} , it must be an intersection point between some face f of $\boldsymbol{\Omega}$ and a hull face of CH(ξ_v) (see Figures 10(b) and (c)) As face f has a portion internal to $CH(\xi_{v})$, which becomes an internal face of $\mathbf{\Omega} \star CH(\xi_v)$, p must be a strong hull vertex of $\mathbf{\Omega} * CH(\xi_v)$. Therefore, all the



Figure 9 Supportable and nonsupportable vertices

computer-aided design



Figure 10 Proof of Lemma 6



Figure 11 Nonsupportable vertex introduced by a nonsupportable edge

vertices of $\Omega * CH(\xi_v)$ are either internal or strong. By Lemma 2, $\Omega * CH(\xi_v)$ is nonconvergent. QED

As an illustration of Lemma 5, vertex v of the pseudopolyhedron in Figure 9(a) is nonsupportable. The regularized intersection between the pseudopolyhedron and $CH(\xi_v)$, where ξ_v are those six vertices adjacent to v, is another pseudopolyhedron, as shown in Figure 9(b), that is nonconvergent By Lemma 5, the ASV series of the pseudopolyhedron in Figure 9(a) does not converge, this can easily be verified

An extension to the supportability of vertices is the supportability of edges. Consider the pseudopolyhedron Ω in Figure 11(a). Its ASV series can easily be shown to be nonconvergent, although all its vertices are supportable

Definition 7. An edge e of a pseudopolyhedron is supportable if there exists a plane containing e such that all the faces incident to e are on one side of that plane, otherwise, edge e is nonsupportable

Lemma 6. If pseudopolyhedron Ω has a nonsupportable edge e, then ASV(Ω) is nonconvergent.

Proof Assume that the nonsupportable edge e has k incident faces f_1, f_2, \ldots, f_k , and that v_1 and v_2 are its two vertices Let p be an arbitrary point on e, but not v_1 or v_2 . Also, let p_i , which is not v_1 or v_2 , be a vertex on face f_i ($i = 1, 2, \ldots, k$) As the line segment [p, p_i] is on a face f_i of Ω , the addition of p to the vertex set of Ω , and the addition of edges [p, v_1], [p, v_2], ..., and [p, p_i] ($i = 1, 2, \ldots, k$) to the edge set of Ω , introduces a new pseudopolyhedron representation of Ω , as

shown in Figure 11(b) As e is nonsupportable, all the points in it, except, possibly, for v_1 or v_2 , are strictly inside CH($\{v_1, v_2, p_1, p_2, ..., p_k\}$). This implies that the vertex p is nonsupportable By Lemma 5, ASV(Ω) is nonconvergent. QED

It should be mentioned that Lemma 5 and Lemma 6 supply a sufficient but not necessary condition for nonconvergence As an example, all the vertices and edges of the polyhedron in Figure 12 are supportable However, its ASV series is nonconvergent

Nevertheless, a linear time algorithm for detecting the sufficiency of nonconvergence offers an attractive alternative to the $O(n^2 \log n)$ time for both necessity and sufficiency

Let o be the origin and (x_1, y_1, z_1) , (x_2, y_2, z_2) , (x_k, y_k, z_k) be k points in the 3D space. If the point o is supportable against $(x_1, y_1, z_1), (x_2, y_2, z_2), \dots$ (x_k, y_k, z_k) , the angle between the normal vector **N**_o of a supporting plane P_o and the vector (x_i, y_i, z_i) must not be greater than 90° for all the i = 1, 2, ..., k (see Figure 13) Conversely, if there exists a vector N_0 such that the angle between it and a vector (x_{ij}, y_{jj}, z_{j}) (i = 1, 2, ...), k) is less than or equal to 90°, then the plane passing through o and orthogonal to N_0 is clearly a supporting plane Therefore, the detection of the supportability becomes the following given k vectors $(x_1, y_1, z_1), (x_2, y_2, z_2), \dots, (x_k, y_k, z_k),$ find another nonzero vector (A, B, C) such that $Ax_i + By_i + Cz_i \ge 0$ (i = 1, 2, ..., k) It is known³ that the solution to this 3-variable problem, if it exists, can be obtained in O(k)time

Let SUPPORT(k, L) be a supportability detection procedure that takes a list L of k points as input, and outputs either 'true', if the origin is supportable against L, or 'false' otherwise With the SUPPORT procedure, the following algorithm is in order.

NSV_DETECT algorithm

Algorit '' De n	hm NSV_DETECT(Ω) etect the existence of nonsupportable vertices in a pseudopolyhedron Ω with vertices
beein	
(1)	tor $i = 1, n, do$
(11)	$v \leftarrow$ the <i>i</i> th vertex in the vertex list V of $\mathbf{\Omega}$
(12)	$\{p_1, p_2, \dots, p_k\} \leftarrow$ those vertices in V that are adjacent to vertex v
(13)	translate $\{p_1, p_2, \dots, p_k\}$ by a displacement of $-v$
(14)	It SUPPORT(k , { p_1, p_2 , p_k }) = 'false' then
	return with 'nonsupportable vertex found
	end if
	end do
(2) end N	return with 'no nonsupportable vertex found' SV_DETECT

To devise an algorithm for detecting the supportability of an edge e of a pseudopolyhedron Ω , let v and v'



Figure 12. Nonconvergent polyhedron with no non-supportable vertices or edges



Figure 13. Angular relationship between the normal of a supporting plane and the adjacent vertices



Figure 14 Finding the point p, on face f,

be the two end points of e, p be its center point, and f_1, f_2, \dots, f_k be the faces of Ω incident to e Also, let p_i be a point on face f_i such that the line segment $[p, p_i]$ completely belongs to f_i ($i = 1, 2, \dots, k$) (see Figure 14). Such a point p_i can be obtained in the constant time from the (clockwise or counterclockwise) order of the edges Let FS(p, f_i) denote the function that returns the point p_i . Referring to the proof of Lemma 6, e is supportable if and only if p is supportable against the point set $\{v, v', p_1, p_2, \dots, p_k\}$. This equivalence relationship gives rise to the following algorithm

NSE_DETECT algorithm

Algorithm NSt_DETECT(Ω) Detect the existence of nonsupportable edges in a pseudopolyhedron Ω with n_i vertices
begin
(1) for $i = 1$ n_e do
(1.1) $y \in V \leftarrow$ the two vertices of the <i>i</i> th edge in the edge list <i>E</i> of Ω
(12) $p \leftarrow$ the center point of $[v, v']$
(1.3) $I_1 I_2 = I_k \leftarrow$ those faces in Ω that are adjacent to the edge $[v, v]$
(1.4) $p_1, p_2, p_1 \leftarrow FS(p, l_1), FS(p, l_2) = FS(p, l_k)$
(15) translate $\{v, v', p_1, p_2, \dots, p_k\}$ by a displacement of $-p$
(1.6) If SUPPORT $(k + 2, \{v, v', p_1, p_2, \dots, p_k\}) = \text{'talse' then}$
return with 'nonsupportable edge found
end it
end do
(2) return with 'no nonsupportable edge tound end NSE_DETECT

Lemma 7 The existence of nonsupportable vertices and nonsupportable edges of a pseudopolyhedron Ω with n_v vertices, n_e edges and n_f faces can be detected in at most $O(n_f)$ time

Proof The theorem is proven by showing that both the NSV_DETECT and NSE_DETECT algorithms run in $O(n_i)$ time For the NSV_DETECT algorithm, because the SUPPORT procedure runs in linear time, the time complexity required by the loop at Step 1 is linear in Σd_i ($i = 1, 2, ..., n_v$), where d_i is the degree of the *i*th vertex in Ω , which has been proven to be $O(n_i)$ in the Appendix of Part 1 of this paper Therefore, NSV_DETECT runs in $O(n_f)$ time. For the NSE_DETECT algorithm, by similar reasoning, the time required by the loop at Step 1 is linear in Σk_i ($i = 1, 2, ..., n_e$), where k_i is the face adjacency index of the *i*th edge in Ω . In the Appendix of Part 1 of this paper, it has been shown that Σk_i ($i = 1, 2, ..., n_e$) is $O(n_f)$. Therefore, NSE_DETECT runs in $O(n_f)$ time.

SUMMARY

It has been established that it takes $O(n^2 \log n)$ time to determine if the ASV series of a given Ω converges In particular, it takes $O(n \log n)$ time to detect if a deficiency Ω , is nonconvergent. To remedy the nonconvergence, an $O(n \log n)$ time algorithm is offered to separate the culprit deficiency Ω , into maximally well connected sets

As an expedient alternative to the $O(n^2 \log n)$ time detection for nonconvergence, the sufficiency condition for nonconvergence can be detected in O(n) time.

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