

# Practical polygonal triangulation in $O(n + k \log k)$ Time

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## Abstract

We present a practical algorithm for triangulating a simple polygon with  $n$  vertices in  $O(n + k \log k)$  time, where  $k$  is the number of local extrema (local maxima, equivalently local minima) with respect to the sweep direction. The algorithm reformulates monotone decomposition to maintain active monotone chains rather than individual edges: the sweep processes only extremal events, while regular vertices are handled via lazy pointer advancement in  $O(n)$  amortized time. Experiments show a **19–46× speedup on convex polygons** over the standard Garey et al. monotone-decomposition baseline. On large random polygons ( $n = 10,000$ ), we achieve **about a 1.3× speedup** over Garey et al.; on star polygons, we observe a **1.2–1.4× speedup** across the tested range.

*Keywords:* simple polygon, triangulation, monotone decomposition, output-sensitive algorithm, plane sweep

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## 1. Introduction

Triangulating simple polygons is a fundamental problem in computational geometry. Chazelle [3] proved that  $O(n)$  time is achievable, but the algorithm’s complexity has limited practical adoption. The plane sweep of Garey et al. [8], running in  $O(n \log n)$  time via monotone decomposition, remains standard; see de Berg et al. [5]. Seidel [13] gave a randomized  $O(n \log^* n)$  expected-time algorithm.

We present an algorithm running in  $O(n + k \log k)$  time, where  $k$  is the number of local extrema (local maxima, equivalently local minima) with respect to the sweep direction. This interpolates between  $O(n)$  for convex polygons ( $k = 1$ ) and  $O(n \log n)$  for worst-case polygons ( $k = \Theta(n)$ ), improving upon classical methods when  $k = o(n / \log n)$ .

The key observation is that the sweep needs to process only local extrema, and regular vertices can be handled implicitly by advancing chain pointers lazily. This yields  $O(k)$  sweep events and  $O(n)$  total pointer advancement, with balanced-tree operations totaling  $O(k \log k)$ .

*Related work.* The triangulation problem has a rich history spanning four decades. The earliest approaches include the ear-clipping method of ElGindy et al. [6], which runs in  $O(n^2)$  time. Garey et al. [8] achieved  $O(n \log n)$  time through monotone decomposition, which remains the standard practical approach; see de Berg et al. [5] for a textbook treatment. Hertel and Mehlhorn [9] provided an alternative  $O(n \log n)$  sweep-line algorithm with improved constant factors.

Subsequent work focused on reducing the  $O(n \log n)$  bound. Tarjan and Van Wyk [14] achieved  $O(n \log \log n)$  using sophisticated data structures, later simplified by Kirkpatrick et al. [12]. Randomized approaches proved fruitful: Clarkson et al. [4] gave an  $O(n \log^* n)$  expected-time Las Vegas algorithm, which Seidel [13] simplified significantly using trapezoidal decomposition. Amato et al. [1] provided further simplifications of randomized linear-time triangulation.

The breakthrough came with Chazelle’s [3] deterministic  $O(n)$  algorithm, resolving a long-standing open problem. However, the algorithm’s complexity—involving hierarchical polygon decomposition and intricate merge operations—has limited practical adoption. Fournier and Montuno [7] studied triangulation of monotone polygons. Keil [10] surveys polygon decomposition more broadly. Output-sensitive algorithms exist for related problems: Kirkpatrick and

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Seidel [11] gave an  $O(n \log h)$  convex hull algorithm where  $h$  is the output size. Chazelle and Incerpi [2] studied shape complexity and triangulation.

Our work contributes an  $O(n + k \log k)$  algorithm that interpolates between these bounds based on polygon complexity. This output-sensitive approach—where  $k$  measures input “difficulty” in terms of sweep events—appears not to have been previously explored for triangulation, despite its naturalness.

*Organization.* Section 2 defines the event-complexity parameter  $k$  and discusses its relationship to reflex vertices. Section 3 presents the algorithm. Section 4 establishes correctness. Section 5 analyzes complexity. Section 6 provides experimental evaluation. Section 7 discusses extensions.

## 2. Preliminaries

Let  $P$  be a simple polygon with vertices  $v_0, v_1, \dots, v_{n-1}$  listed in counterclockwise order along the boundary  $\partial P$ . We write  $v_i = (x_i, y_i)$  for the coordinates of each vertex and adopt the convention that indices are taken modulo  $n$ , so  $v_{-1} = v_{n-1}$  and  $v_n = v_0$ . The *interior angle* at vertex  $v_i$  is the angle  $\angle v_{i-1}v_iv_{i+1}$  measured inside  $P$ . A vertex is *convex* if its interior angle is at most  $\pi$  and *reflex* if its interior angle strictly exceeds  $\pi$ . We denote by  $r$  the number of reflex vertices.

**Definition 1** (Vertex classification). Assuming general position (no two vertices share the same  $y$ -coordinate), each vertex  $v_i$  is classified according to the relative  $y$ -coordinates of its neighbors:

- **Start vertex:**  $y_{i-1} < y_i$  and  $y_{i+1} < y_i$ , with interior angle  $< \pi$ .
- **Split vertex:**  $y_{i-1} < y_i$  and  $y_{i+1} > y_i$ , with interior angle  $> \pi$ .
- **End vertex:**  $y_{i-1} > y_i$  and  $y_{i+1} > y_i$ , with interior angle  $< \pi$ .
- **Merge vertex:**  $y_{i-1} > y_i$  and  $y_{i+1} < y_i$ , with interior angle  $> \pi$ .
- **Regular vertex:** exactly one neighbor has  $y$ -coordinate greater than  $y_i$ .

Start and split vertices are *local maxima*; end and merge vertices are *local minima*. Split and merge vertices are precisely the reflex vertices among local extrema. Regular vertices partition into two subtypes based on whether the polygon interior lies to their left or right as one traverses the boundary; this distinction, while important for implementation, does not affect our analysis.

**Definition 2** (Monotone chain). A *y-monotone chain* is a maximal contiguous sequence of boundary vertices  $v_a, v_{a+1}, \dots, v_b$  such that the  $y$ -coordinates are strictly monotonic (either strictly increasing or strictly decreasing) along the sequence. Each chain connects a local maximum to a local minimum.

The boundary  $\partial P$  decomposes uniquely into monotone chains, with consecutive chains sharing their endpoint extrema. If there are  $k$  local maxima, there are also  $k$  local minima (since the boundary is a closed curve), and hence  $2k$  chains.

**Remark 3** (Event-complexity parameter  $k$ ). Our running time and analysis are stated in terms of  $k$ , the number of local extrema (equivalently, local maxima). This parameter directly controls the number of sweep events in the chain-based formulation.

## 3. Algorithm

The algorithm consists of three phases: (1) chain construction and vertex classification, (2) monotone decomposition via chain-based plane sweep, and (3) triangulation of monotone pieces. We describe each phase in detail.

### 3.1. Phase 1: Chain Construction

A single traversal of  $\partial P$  classifies each vertex according to Definition 1 and partitions the boundary into monotone chains. For each vertex  $v_i$ , we compare  $y_i$  to  $y_{i-1}$  and  $y_{i+1}$  and compute the cross product  $(v_{i+1} - v_i) \times (v_i - v_{i-1})$  to determine convexity. Simultaneously, we record each chain as an array of vertex indices from its upper endpoint (local maximum) to its lower endpoint (local minimum). For each local minimum  $v$ , we store a pointer to the unique *left-boundary chain* terminating at  $v$ —the chain for which, when traversed from upper to lower endpoint, the polygon interior lies to the right.

**Definition 4** (Left-boundary chain). A monotone chain is a *left-boundary chain* if, when traversed from its upper endpoint to its lower endpoint, the polygon interior lies to the right of the traversal direction. Equivalently, for a counterclockwise-oriented polygon, a chain is left-boundary if its traversal direction (downward) opposes the boundary orientation.

At each local maximum, exactly one of the two originating chains is left-boundary; at each local minimum, exactly one of the two terminating chains is left-boundary. This phase runs in  $O(n)$  time and produces  $O(k)$  chains, where  $k$  is the number of local maxima (equivalently local minima).

### 3.2. Phase 2: Monotone Decomposition

A polygon is *y-monotone* if every horizontal line intersects it in a connected set (either empty, a point, or a segment). Split vertices violate monotonicity by creating local maxima where the boundary diverges downward; merge vertices violate it by creating local minima where boundary paths converge from above. The decomposition phase inserts diagonals to eliminate all split and merge vertices, partitioning  $P$  into *y-monotone* subpolygons.

*Sweep-line status structure.* Unlike the classical algorithm, which maintains individual edges in a balanced search tree  $T$ , we maintain *active left-boundary chains*. A chain  $C$  is *active* at sweep height  $y$  if  $y$  lies strictly between the  $y$ -coordinates of  $C$ 's upper and lower endpoints. The tree  $T$  stores active left-boundary chains ordered by their  $x$ -coordinate at the current sweep height.

Each chain  $C$  in  $T$  maintains:

1. **Edge pointer**  $C.curr$ : initialized to the topmost edge of  $C$ , this pointer tracks our position within the chain. The upper vertex of  $C.curr$  serves as the *slab entry*—the default diagonal target when no pending merge exists.
2. **Pending merge**  $C.pending$ : either null or a merge vertex awaiting connection to a lower vertex. When a merge vertex  $v$  is processed and  $C$  is immediately to  $v$ 's left, we set  $C.pending \leftarrow v$ .

*Lazy edge pointer advancement.* The comparison function for  $T$ , when comparing chains at sweep height  $y$ , first advances each chain's edge pointer to ensure the current edge spans  $y$ :

```

1: procedure ADVANCE( $C, y$ )
2:   while  $C.curr.lower.y > y$  do
3:      $C.curr \leftarrow$  next edge down  $C$ 
4:   end while
5: end procedure

```

After advancement, the  $x$ -coordinate of  $C$  at height  $y$  is computed by linear interpolation along  $C.curr$ . Since each vertex is visited by at most one chain's pointer exactly once, the total cost of all pointer advancements is  $O(n)$ , amortized over all tree operations.

*Event processing.* The sweep processes local extrema in decreasing  $y$ -order. Let  $E$  denote the sorted list of extrema;  $|E| = 2k$ , where  $k$  is the number of local maxima (equivalently local minima).

The complete decomposition procedure is given in Algorithm 1. For split vertices, we connect upward to either a pending merge (if one exists on the immediately-left chain) or to the slab entry (the upper vertex of the current edge on that chain). For merge vertices, we first resolve any pending merges on both the terminating chain and the chain to its left, then register the current vertex as pending on the left chain.

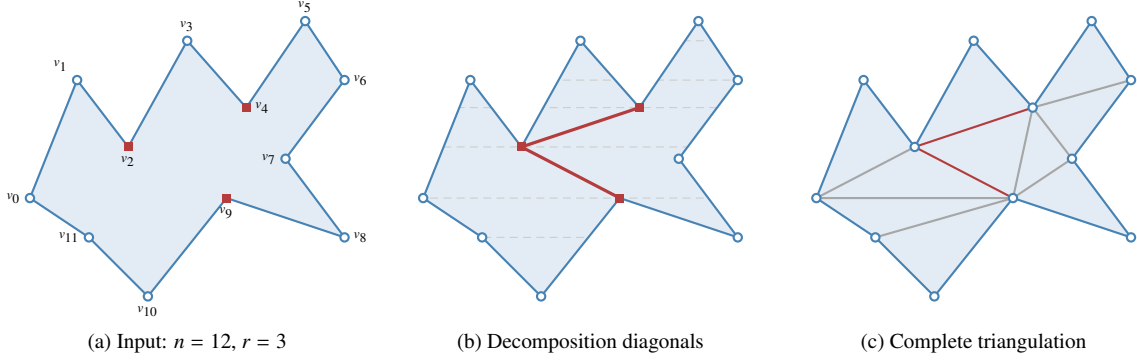


Figure 1: Triangulation of a 12-vertex polygon with 3 reflex vertices (squares). (a) Input polygon  $P$ ; circles denote convex vertices, squares denote reflex vertices  $v_2$  (merge),  $v_4$  (merge),  $v_9$  (split). (b) Monotone decomposition: dashed lines indicate sweep heights at local extrema; bold diagonals  $(v_2, v_4)$  and  $(v_9, v_2)$  partition  $P$  into three  $y$ -monotone pieces. (c) Final triangulation with  $n - 2 = 10$  triangles; decomposition diagonals in bold, triangulation diagonals in gray.

### 3.3. Phase 3: Triangulation

The diagonals from Phase 2 partition  $P$  into  $y$ -monotone subpolygons. We construct a doubly-connected edge list (DCEL) or equivalent adjacency structure from the original boundary edges plus the  $|D| \leq r$  diagonals. Each face is extracted by traversing half-edges, yielding the vertex sequence of each monotone subpolygon.

Each  $y$ -monotone polygon with  $m$  vertices is triangulated in  $O(m)$  time using the classical stack-based algorithm: vertices are processed in  $y$ -sorted order, maintaining a stack representing the “reflex chain” of vertices not yet triangulated. When a vertex from the opposite chain is encountered, all stack vertices are triangulated; when a vertex from the same chain is encountered, we triangulate as many stack vertices as remain visible. Since the sum of face sizes equals  $n + 2|D| = O(n)$ , the total triangulation time is  $O(n)$ .

### 3.4. Illustrative Example

Figure 1 illustrates the algorithm on a polygon with  $n = 12$  vertices and  $r = 3$  reflex vertices ( $v_2, v_4$  are merges;  $v_9$  is a split). The sweep processes 9 local extrema. At merge  $v_4$ , the pending mechanism registers  $v_4$  on the left chain. At merge  $v_2$ , the pending  $v_4$  triggers diagonal  $(v_2, v_4)$ , then  $v_2$  becomes pending. At split  $v_9$ , the pending  $v_2$  triggers diagonal  $(v_9, v_2)$ . The two diagonals partition  $P$  into three  $y$ -monotone faces, yielding 10 triangles.

## 4. Correctness

The geometric idea of Algorithm 1 is to implement the classical monotone-decomposition sweep (e.g., de Berg et al. [5]) while exploiting the fact that regular vertices occur only along monotone chains. Instead of processing every regular vertex as an explicit event, we maintain, for each active chain, a pointer to its current edge at the sweep height and update it lazily. Correctness is therefore best stated as an equivalence to the textbook sweep, rather than via ad-hoc visibility regions.

*Textbook baseline (edge-based sweep).* The classical algorithm maintains a balanced BST of active edges intersecting the sweep line, ordered by their intersection  $x$ -coordinate. Each active edge  $e$  stores a vertex *helper*( $e$ ), defined as the most recently processed vertex above the sweep line that can “see”  $e$  from the interior; diagonals are inserted from split/merge vertices to the helper of the edge immediately to their left. The resulting diagonal set is known to be valid and to yield a  $y$ -monotone decomposition; see [5].

*Chain-based representation.* Fix a sweep height  $y$  between two consecutive event levels (no vertex has  $y$ -coordinate exactly  $y$ ). For a chain  $C$  active at height  $y$ , let  $e(C, y)$  denote the unique edge of  $C$  whose  $y$ -span contains  $y$  (after ADVANCE). Our BST  $T$  stores active left-boundary chains ordered by the  $x$ -coordinate of  $e(C, y)$  at height  $y$ ; this is well-defined under general position.

*Tie-breaking and degeneracies..* Throughout, we assume general position (no two vertices share the same  $y$ -coordinate) so that event levels are distinct and each non-extremal vertex is strictly above one neighbor and strictly below the other. If equal  $y$ -coordinates are allowed, a standard symbolic perturbation (or lexicographic order by  $(y, x)$  together with corresponding non-strict inequalities in Definition 1) can be used; this does not affect the algorithmic structure and is omitted for clarity. We also evaluate  $e(C, y)$  at a height  $y$  strictly between event levels, so no sweep query is performed exactly at a vertex height.

**Definition 5** (Implicit helper state). For an active chain  $C$  at sweep height  $y$ , define its *implicit helper vertex*

$$ih(C, y) := \text{the upper endpoint of } e(C, y).$$

Intuitively,  $ih(C, y)$  is the last vertex of  $C$  that the sweep has passed strictly above height  $y$ .

**Lemma 6** (Advance simulates regular-vertex helper updates). *Fix any chain  $C$  and consider two sweep heights  $y_1 > y_2$  with no extrema between them. Let  $e_1 = e(C, y_1)$  and  $e_2 = e(C, y_2)$  after performing ADVANCE as needed. Then  $ih(C, y_2)$  equals the last vertex of  $C$  encountered when walking down  $C$  from its top endpoint until reaching height  $y_2$ ; moreover, along this descent, every regular vertex of  $C$  is encountered exactly once over the entire sweep.*

*Proof.* By definition, ADVANCE walks the chain pointer  $C.curr$  monotonically downward along  $C$ , stopping once the current edge spans the query height. Therefore the upper endpoint of the resulting current edge is exactly the last chain vertex above the height, i.e.,  $ih(C, y_2)$ . Since  $C.curr$  only moves forward (down the chain), each vertex of  $C$  can become an upper endpoint of  $C.curr$  at most once, implying the claimed “encountered once” property.  $\square$

**Definition 7** (Pending as “merge-helper”). In the textbook sweep, the only helper values that trigger a diagonal later are merge vertices. We encode this by storing, for each active chain  $C$ , a field  $C.pending$  which is either null or a merge vertex. We interpret

$$C.pending \neq \text{NULL} \iff \text{helper}(e(C, y)) \text{ is a merge vertex, and equals } C.pending.$$

If  $C.pending = \text{NULL}$ , the role of  $\text{helper}(e(C, y))$  is played by the implicit helper  $ih(C, y)$  from Definition 5.

**Lemma 8** (Equivalence of diagonal targets). *At every event vertex  $v$  (start/end/split/merge), the diagonal(s) added by Algorithm 1 coincide with those added by the textbook edge-based sweep when run on the same polygon and event order.*

*Proof.* We compare Algorithm 1 to the textbook sweep run on the same polygon, with the same decreasing- $y$  event order.

**Step 1: Ordering agreement.** Fix any height  $y$  strictly between consecutive event levels. For every active chain  $C$ , the edge  $e(C, y)$  intersects the sweep line in exactly one point. By definition of our BST keys, the in-order ordering of chains in  $T$  is the left-to-right ordering of these intersection points, hence agrees with the textbook ordering of the corresponding active edges. Therefore, when Algorithm 1 takes “the predecessor of  $v$  in  $T$ ”, it identifies the same geometric object as “the active edge immediately left of  $v$ ” in the textbook sweep.

**Step 2: Helper agreement on regular vertices via ADVANCE.** The textbook sweep updates  $\text{helper}(e)$  at regular vertices when the current active edge on a chain changes to the next edge below. Our algorithm does not process those regular vertices as events, but it does update the current edge implicitly: whenever  $T$  needs the key of a chain at some height  $y$ , it calls ADVANCE so that the stored edge equals  $e(C, y)$ . By Lemma 6, the vertex  $ih(C, y)$  equals the last vertex encountered when walking down the chain to height  $y$ , which is exactly the vertex that the textbook sweep would have last assigned as helper on the current edge of that chain unless that helper is a merge vertex awaiting a future diagonal. We encode precisely this exception via Definition 7.

**Step 3: Case analysis at event vertices.** Let  $v$  be the next event vertex processed.

- **Start.** In the textbook sweep, a start vertex inserts its outgoing edge(s) into the status structure and sets their helpers to  $v$ . In our representation, the unique left-boundary chain originating at  $v$  is inserted into  $T$  and its pointer is initialized so that  $ih(C, y) = v$  for heights just below  $v$ ; also  $C.pending$  is null. Thus the (merge-)helper information agrees.

- **End.** In the textbook sweep, the edge terminating at an end vertex is removed, and if its helper is a merge vertex then a diagonal is added from  $v$  to that helper. In our representation, we remove the terminating left-boundary chain  $R$ ; if  $R.pending \neq \text{NULL}$  we add the same diagonal before removal. Hence the diagonal action agrees.
- **Split.** In the textbook sweep, at a split vertex  $v$  one finds the active edge  $e$  immediately left of  $v$  and adds a diagonal from  $v$  to  $helper(e)$ ; then  $helper(e)$  is set to  $v$ , and the new edge leaving  $v$  is inserted with helper  $v$ . In our representation,  $L$  is the predecessor chain (Step 1), so its current edge equals  $e(L, v.y)$ ; by Step 2, the helper used by the textbook sweep on that edge is either  $L.pending$  (if it is a merge) or  $ih(L, v.y) = L.curr.upper$  (otherwise). Algorithm 1 adds exactly the corresponding diagonal, clears  $L.pending$  if it was used, and inserts the new left-boundary chain from  $v$  with implicit helper  $v$ . Thus the diagonal target agrees.
- **Merge.** In the textbook sweep, at a merge vertex  $v$  one may add a diagonal to the helper of the edge terminating at  $v$  if that helper is a merge, removes that edge, then finds the edge immediately left of  $v$  and may add a diagonal to its merge helper, finally setting the helper of the left edge to  $v$ . In our representation, we (i) remove the chain terminating at  $v$  after emitting a diagonal to its pending merge (if any), (ii) emit a diagonal to the pending merge of the immediately-left chain (if any), and (iii) set that chain's pending to  $v$ . Therefore the diagonal targets agree.

□

**Theorem 9** (Correctness). *Algorithm 1 produces a valid set of non-crossing diagonals partitioning  $P$  into  $y$ -monotone subpolygons.*

*Proof.* By Lemma 8, Algorithm 1 outputs exactly the same diagonals as the textbook monotone-decomposition sweep, interpreted through our chain representation. The textbook sweep's diagonals are interior, non-crossing, and remove all split/merge vertices, yielding  $y$ -monotone pieces [5]. Hence the same properties hold for our output. □

## 5. Complexity Analysis

**Lemma 10** (Diagonal bound). *Let  $s$  and  $m$  be the numbers of split and merge vertices of  $P$  (equivalently, the numbers of reflex local maxima and reflex local minima). Then Algorithm 1 inserts at most  $s + m$  diagonals. In particular,  $|D| \leq s + m \leq r$ .*

*Proof.* Each split vertex triggers exactly one diagonal insertion in Algorithm 1, contributing  $s$  diagonals.

It remains to bound diagonals created due to the pending mechanism. Only merge vertices are ever assigned to a field  $C.pending$ . Each such assignment  $C.pending \leftarrow v$  occurs at the unique processing step of merge vertex  $v$  and can happen only once for that vertex. Afterwards, that stored value is consumed exactly once: either it is used to form a diagonal at a later split/merge event when the chain is immediately left of the current vertex, or it is used at the end vertex where the chain terminates; in all cases the stored value is overwritten (or the chain is removed), so it cannot generate a second diagonal. Therefore at most one diagonal is generated per merge vertex via pending, contributing at most  $m$  diagonals.

Hence  $|D| \leq s + m \leq r$  since every split and merge vertex is reflex. □

**Theorem 11** (Complexity). *A simple polygon with  $n$  vertices and  $k$  local maxima (equivalently  $k$  local minima) can be triangulated in  $O(n + k \log k)$  time and  $O(n)$  space.*

*Proof.* **Phase 1 (Chain construction):** A single traversal classifies all vertices and constructs all chains in  $O(n)$  time using  $O(n)$  space.

**Phase 2 (Monotone decomposition):**

- **Sorting:** The  $2k$  local extrema are sorted in  $O(k \log k)$  time.
- **Event processing:** There are  $2k$  events. Each event involves  $O(1)$  BST operations (insertions, deletions, predecessor queries), each taking  $O(\log k)$  time since  $|T| = O(k)$ .

- *Edge pointer advancement*: The comparison function advances edge pointers lazily. Each of the  $n$  vertices is visited at most once across all advancements (each vertex belongs to exactly one chain and is passed exactly once by that chain’s pointer). Total advancement cost:  $O(n)$ .

The decomposition phase totals  $O(n + k \log k)$  time.

*Phase 3 (Triangulation)*: Constructing the adjacency structure takes  $O(n + |D|) = O(n)$  time by Lemma 10. Face extraction and triangulation together take  $O(\sum_f |f|)$  time, where the sum is over all faces  $f$ . Since faces partition the plane inside  $P$ , and each original edge and diagonal appears in exactly two face boundaries,  $\sum_f |f| = 2(n + |D|) = O(n)$ .

*Total time*:  $O(n) + O(n + k \log k) + O(n) = O(n + k \log k)$ .

*Space*: Storing the polygon requires  $O(n)$  space. The chain data structures use  $O(n)$  total (chains partition the vertices). The BST  $T$  contains at most  $O(k)$  chains, each with  $O(1)$  auxiliary data. The adjacency structure for face extraction uses  $O(n + |D|) = O(n)$  space. Total:  $O(n)$ .  $\square$

*Constant factors in the  $k \log k$  term.* To compare practical performance between output-sensitive sweeps, it is useful to separate (i) the number of balanced-tree operations from (ii) the cost of an order comparison in the tree. In our method, the balanced structure  $T$  stores only *active chains* and is queried only at local extrema and split/merge vertices. Consequently, the number of tree operations is *linear in  $k$* : each event performs one predecessor query in  $T$  and  $O(1)$  updates (insert/delete of at most a constant number of chains), and there are exactly  $2k$  events. Thus, the total number of tree operations is bounded by  $c_T k + O(1)$  for a small constant  $c_T$ . Each tree operation performs  $O(\log k)$  key comparisons.

The key comparison between two active chains is where our constant-factor savings arise. In a classical edge-based sweep (as in Hertel–Mehlhorn [9]), the ordering predicate evaluates the  $x$ -coordinate of each active edge at the current sweep level  $y$ , which involves arithmetic on the two edge endpoints (in particular, at least one division per evaluation). In our chain representation, comparisons advance chain pointers *lazily*: when the sweep level descends, a chain pointer moves only forward along the boundary, and each boundary vertex is advanced at most once over the entire run. This turns the potentially expensive re-evaluation of edge intersections into an amortized  $O(1)$  pointer update stream, keeping the constant in front of  $k \log k$  dominated by the tree’s comparison count rather than repeated geometric recomputation.

**Corollary 12.** *For  $k = o(n / \log n)$ , the algorithm runs in  $o(n \log n)$  time.*

## 6. Experimental Evaluation

We implemented the algorithm in C++ and compared against: Garey et al. [8] (monotone decomposition via the `polypartition` library), Hertel–Mehlhorn [9] (convex partition), and Seidel [13] (incremental randomized triangulation). All experiments were run with GCC at -O3. To prevent pathological slowdowns from dominating wall-clock time, each baseline run was capped with a fixed timeout; missing entries in tables are shown as “–”. Our implementation uses a hybrid strategy: it runs the chain-based sweep when  $k$  is small (we use a simple threshold  $k < \max(8, n/8)$ ), and otherwise falls back to an edge-based sweep with a linked representation for improved constants on high-event inputs.

*Benchmark setup.* We report *mean  $\pm$  standard deviation* over `polygons_per_config=5` instances for each (family,  $n$ ) configuration (seeds  $0, \dots, 4$ ). All polygon families are generated deterministically from the seed and  $n$ , followed by a fixed rotation to avoid accidental equal- $y$  degeneracies. The `RANDOM` family uses the same generator as `scripts/generate_polygons`: vertices are sampled by sorting random angles and assigning random radii (yielding a simple star-shaped polygon). The `CONVEX` family uses seeded affine images of a regular  $n$ -gon; `DENT` pulls one seeded vertex inward (seeded location and depth); and `STAR` uses a seeded phase and radii (with mild jitter). The event parameter  $k$  shown in the tables is computed directly from each polygon as the number of local maxima, and is reported as mean  $\pm$  stdev across instances.

Our main random-polygon results are included verbatim from the benchmark harness:

For convex inputs, our implementation benefits from the linear-time fan fast path and achieves a **19–46 $\times$  speedup** over Garey et al. in our suite. For star polygons, we observe a **1.2–1.4 $\times$  speedup** across the tested range (about **1.4 $\times$**  at  $n = 10,000$ ). For random polygons, we achieve **about a 1.3 $\times$  speedup** over Garey et al. at  $n = 10,000$  and are typically

Table 1: Running time (ms) on random polygons (mean  $\pm$  stdev over instances).

$n$	$k$	<b>Ours</b>	Seidel	Garey	Hertel–Mehlhorn
100	$29 \pm 3$	<b><math>0.04 \pm 0.00</math></b>	$6.86 \pm 0.57$	$0.05 \pm 0.01$	$0.18 \pm 0.01$
500	$160 \pm 2$	<b><math>0.19 \pm 0.05</math></b>	$7.42 \pm 0.75$	$0.22 \pm 0.01$	$2.64 \pm 0.11$
1,000	$329 \pm 4$	<b><math>0.34 \pm 0.01</math></b>	$7.56 \pm 0.68$	$0.44 \pm 0.04$	$9.63 \pm 0.09$
2,000	$666 \pm 9$	<b><math>0.73 \pm 0.09</math></b>	$8.96 \pm 1.04$	$1.08 \pm 0.24$	$42.06 \pm 3.60$
5,000	$1661 \pm 17$	<b><math>2.06 \pm 0.34</math></b>	$12.01 \pm 1.14$	$2.35 \pm 0.17$	$246.78 \pm 8.14$
10,000	$3316 \pm 12$	<b><math>3.90 \pm 0.54</math></b>	$17.80 \pm 0.94$	$5.03 \pm 0.64$	$1178.57 \pm 78.66$

faster across the tested range. On dent polygons (very small  $k$ ), we are competitive at small  $n$  and become faster as  $n$  grows. Our method significantly outperforms Hertel–Mehlhorn (which times out at large  $n$  on star polygons); Seidel exhibits high constant factors in our experiments.

## 7. Discussion

*Comparison with existing algorithms.* Table 2 summarizes theoretical complexity. The primary comparison is against the classical  $O(n \log n)$  monotone decomposition of Garey et al. [8] and Seidel’s  $O(n \log^* n)$  randomized algorithm [13].

Table 2: Algorithm comparison.

Algorithm	Complexity	Notes
Garey et al. [8]	$O(n \log n)$	standard practical
Hertel–Mehlhorn [9]	$O(n \log n)$	convex partition
Seidel [13]	$O(n \log^* n)$ exp.	randomized
Chazelle [3]	$O(n)$	impractical
<b>This paper</b>	$O(n + k \log k)$	output-sensitive (in #extrema $k$ )

Our algorithm achieves the output-sensitive bound  $O(n + k \log k)$ , where  $k$  is the number of local extrema with respect to the sweep direction. The sweep processes  $O(k)$  events with  $O(\log k)$  balanced-tree operations each, while lazy pointer advancement handles all  $n$  vertices in  $O(n)$  amortized time. This interpolates between  $O(n)$  on nearly monotone inputs and  $O(n \log n)$  in the worst case. The implementation uses cache-friendly monotone chain representation for efficient constant factors.

*Extensions.* The algorithm extends naturally to polygons with holes: each hole boundary contributes its own set of chains and extrema, and the sorted extrema lists are merged. The analysis carries through with  $k$  now denoting the total number of local maxima across all boundaries (equivalently, local minima).

*Implementation.* For  $k$  close to  $n$ , a simpler edge-based sweep may have better constants. A hybrid approach using chains when  $k < n / \log n$  balances asymptotic and practical performance.

*Open problems.* Can the  $O(k \log k)$  term be reduced to  $O(k)$ ? Do analogous bounds hold for polygons with holes or higher-dimensional tetrahedralization?

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Table A.3: Running time (ms) across polygon families (mean over instances).

Type	$n$	$k$	Ours	Seidel	Garey	Hertel–Mehlhorn
Convex	100	$1 \pm 0$	<b>0.00</b> $\pm$ <b>0.00</b>	$6.44 \pm 0.50$	$0.03 \pm 0.01$	$0.00 \pm 0.00$
	500	$1 \pm 0$	<b>0.00</b> $\pm$ <b>0.00</b>	$6.22 \pm 0.26$	$0.11 \pm 0.00$	$0.01 \pm 0.00$
	1,000	$1 \pm 0$	<b>0.02</b> $\pm$ <b>0.02</b>	$7.31 \pm 0.90$	$0.22 \pm 0.05$	$0.03 \pm 0.00$
	2,000	$1 \pm 0$	<b>0.02</b> $\pm$ <b>0.01</b>	$8.41 \pm 0.68$	$0.48 \pm 0.02$	$0.05 \pm 0.01$
	5,000	$1 \pm 0$	<b>0.06</b> $\pm$ <b>0.00</b>	$9.99 \pm 1.04$	$1.20 \pm 0.03$	$0.14 \pm 0.00$
	10,000	$1 \pm 0$	<b>0.15</b> $\pm$ <b>0.05</b>	$14.81 \pm 2.79$	$3.07 \pm 0.52$	$0.33 \pm 0.05$
Dent	100	$2 \pm 0$	$0.04 \pm 0.00$	$7.56 \pm 0.44$	<b>0.03</b> $\pm$ <b>0.00</b>	$0.44 \pm 0.04$
	500	$2 \pm 0$	$0.15 \pm 0.09$	$7.23 \pm 0.31$	<b>0.15</b> $\pm$ <b>0.04</b>	$35.29 \pm 4.70$
	1,000	$2 \pm 0$	$0.24 \pm 0.03$	$7.10 \pm 0.77$	<b>0.23</b> $\pm$ <b>0.03</b>	$745.27 \pm 975.52$
	2,000	$2 \pm 0$	<b>0.42</b> $\pm$ <b>0.10</b>	$9.05 \pm 2.86$	$0.52 \pm 0.11$	$2346.05 \pm 472.53$
	5,000	$2 \pm 0$	<b>0.96</b> $\pm$ <b>0.13</b>	$9.46 \pm 1.22$	$1.15 \pm 0.16$	–
	10,000	$2 \pm 0$	<b>1.97</b> $\pm$ <b>0.27</b>	$16.48 \pm 2.47$	$3.00 \pm 0.60$	–
Random	100	$29 \pm 3$	<b>0.04</b> $\pm$ <b>0.00</b>	$6.86 \pm 0.57$	$0.05 \pm 0.01$	$0.18 \pm 0.01$
	500	$160 \pm 2$	<b>0.19</b> $\pm$ <b>0.05</b>	$7.42 \pm 0.75$	$0.22 \pm 0.01$	$2.64 \pm 0.11$
	1,000	$329 \pm 4$	<b>0.34</b> $\pm$ <b>0.01</b>	$7.56 \pm 0.68$	$0.44 \pm 0.04$	$9.63 \pm 0.09$
	2,000	$666 \pm 9$	<b>0.73</b> $\pm$ <b>0.09</b>	$8.96 \pm 1.04$	$1.08 \pm 0.24$	$42.06 \pm 3.60$
	5,000	$1661 \pm 17$	<b>2.06</b> $\pm$ <b>0.34</b>	$12.01 \pm 1.14$	$2.35 \pm 0.17$	$246.78 \pm 8.14$
	10,000	$3316 \pm 12$	<b>3.90</b> $\pm$ <b>0.54</b>	$17.80 \pm 0.94$	$5.03 \pm 0.64$	$1178.57 \pm 78.66$
Star	100	$49 \pm 1$	<b>0.04</b> $\pm$ <b>0.00</b>	$6.70 \pm 0.55$	$0.05 \pm 0.00$	$0.18 \pm 0.02$
	500	$247 \pm 1$	<b>0.16</b> $\pm$ <b>0.01</b>	$7.05 \pm 0.10$	$0.21 \pm 0.03$	$6.64 \pm 0.32$
	1,000	$498 \pm 1$	<b>0.30</b> $\pm$ <b>0.01</b>	$7.25 \pm 0.07$	$0.41 \pm 0.04$	$42.44 \pm 0.75$
	2,000	$998 \pm 1$	<b>0.55</b> $\pm$ <b>0.03</b>	$8.07 \pm 0.78$	$0.78 \pm 0.04$	$378.96 \pm 43.19$
	5,000	$2499 \pm 1$	<b>1.61</b> $\pm$ <b>0.19</b>	$10.52 \pm 1.10$	$2.01 \pm 0.28$	$5618.17 \pm 1044.87$
	10,000	$4998 \pm 1$	<b>3.41</b> $\pm$ <b>0.56</b>	$17.15 \pm 0.94$	$4.27 \pm 0.41$	–

## Appendix A. Extended Benchmark Results

The full results across all polygon families appear below (generated directly from the benchmark harness):

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**Algorithm 1** Chain-Based Monotone Decomposition

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**Require:** Polygon  $P$  with classified vertices and identified left-boundary chains

**Ensure:** Diagonal set  $D$  partitioning  $P$  into  $y$ -monotone subpolygons

```
1:  $E \leftarrow$  local extrema sorted by decreasing  $y$ -coordinate
2:  $T \leftarrow$  empty balanced BST of active left-boundary chains
3:  $D \leftarrow \emptyset$ 
4: for each extremum  $v$  in  $E$  do
5:   switch type of  $v$ 
6:   case Start
7:     Insert left-boundary chain originating at  $v$  into  $T$ 
8:     Initialize  $C.curr$  to top edge,  $C.pending \leftarrow \text{NULL}$ 
9:   case End
10:     $R \leftarrow$  left-boundary chain terminating at  $v$ 
11:    if  $R.pending \neq \text{NULL}$  then
12:       $D \leftarrow D \cup \{(v, R.pending)\}$  ▷ Connect pending merge downward
13:    end if
14:    Remove  $R$  from  $T$ 
15:   case Split
16:     $L \leftarrow$  predecessor of  $v$  in  $T$  ▷ Chain immediately left of  $v$ 
17:    if  $L.pending \neq \text{NULL}$  then
18:       $D \leftarrow D \cup \{(v, L.pending)\}$ ;  $L.pending \leftarrow \text{NULL}$ 
19:    else
20:       $D \leftarrow D \cup \{(v, L.curr.upper)\}$  ▷ Slab entry
21:    end if
22:    Insert left-boundary chain originating at  $v$  into  $T$ 
23:   case Merge
24:     $R \leftarrow$  left-boundary chain terminating at  $v$ 
25:     $L \leftarrow$  predecessor of  $R$  in  $T$  ▷ Chain immediately left
26:    if  $R.pending \neq \text{NULL}$  then
27:       $D \leftarrow D \cup \{(v, R.pending)\}$ 
28:    end if
29:    if  $L.pending \neq \text{NULL}$  then
30:       $D \leftarrow D \cup \{(v, L.pending)\}$ 
31:    end if
32:     $L.pending \leftarrow v$ 
33:    Remove  $R$  from  $T$ 
34: end for
35: return  $D$ 
```

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