

Emumerating small hyperbolic 3-manifolds

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0. Introduction

This paper documents a rigorous computer-assisted procedure for analyzing hyperbolic 3-manifolds.

The first two applications of this procedure was published as part of [GMT], as Propositions 1.28 and 2.8. The proof of Proposition 1.28 is reproduced here with only minor changes, including most of Section 1 and the entirety of Sections 2 through 5. This document introduces techniques useful for "finding killerwords" in Section 6, and in so doing borrows from material that was originally published as remarks in [GMT].

k that every hyperbolic 3-manifold has a non-coalescible insulator family, and therefore that homotopy hyperbolic 3-manifolds are hyperbolic. Formally

Informally, the result was first published as Conjecture 1.31 of [GMT], and recently become theorem:

THEOREM 0.1. *Let M be an orientable hyperbolic 3-manifold, and let δ be a shortest geodesic. Then, either $tuberadius(\delta) > \ln(3)/2$, or M is isometric to one of N_0, \dots, N_6 , where:*

- i) N_0 is Vol3.
- ii) N_1 is FIXME.
- iii) N_2 is FIXME.
- iv) N_3 is FIXME.
- v) N_4 is FIXME.
- vi) N_5 is FIXME.
- vii) N_6 is FIXME.

Proof. Let X_0, \dots, X_6 denote the exceptional regions of Theorem 2.XX, with N_i the corresponding conjectural manifold. [GMT] showed that N_0 =Vol3 is the unique manifold in the region and [JR] showed that Vol3 covered no

3-manifold. [JR] also proved that N_5 and N_6 are isometric. [CLLMR] and [L] identify a manifold in each region and [CLLMR] show that these manifolds are the unique ones in its region. [CLLMR] show that N_1 , N_5 and N_6 cover no manifold. In [GT] the proof is completed by showing that N_3 covers no manifold and each of N_2 and N_4 2-fold cover manifolds, however the quotients are all non exceptional, i.e. each shortest geodesic has a $\log(3)/2$ tube. \square

Other applications of the procedure:

Theorem [G] (Smale conjecture of hyperbolic 3-manifolds) If N is a closed hyperbolic 3-manifold, then the natural inclusion $Isom(N) \rightarrow Diff(N)$ is a homotopy equivalence.

The proof made essential use of the fact that a shortest geodesic of a closed hyperbolic 3-manifold satisfies the insulator condition [GMT]

Theorem [GMM], [Mi] The Week's manifold is the unique closed orientable hyperbolic 3-manifold of minimal volume.

This result relies on

Theorem [ACS] Suppose that M is a closed, orientable hyperbolic 3-manifold, and that C is a shortest geodesic in M . Set $N = D \text{ drill } C \cdot M / \sim$. If $tub_{rad}(C) \geq \log 3/2$ then $vol N \leq 3.0177 \cdot vol M$.

This is based on a result of [ADST], that makes essential use of Perelman's [Pe] monotonicity result, a key element of his proof of geometrization and the $\log(3)/2$ theorem of [GMT]. Theorem C is used to prove other results such as

Theorem [ACS] Suppose that M is a closed, orientable, hyperbolic 3-manifold such that $vol(M) \leq 1.22$. Then $H_1(M : \mathbb{Z}p)$ has dimension at most 3 for every prime p .

(The statement is sharpened slightly to take into account [GT])

Theorem [LM] Let M be a compact orientable 3-manifold with boundary a torus, and with interior admitting a complete finite-volume hyperbolic structure. Then the number of exceptional slopes on M is at most 10.

This result, known as the Gordon conjecture, relies on rigorous computer assistance using the AffApprox technology.

Finally, the procedure is currently in use by Gabai, Thurston, and Yarmola to rigorously analyze small orientable cusped hyperbolic 3-manifolds, and to rigorously analyze small non-orientable closed hyperbolic 3-manifolds.

* ORGANIZATION This paper is organized as follows.

In Section 1 we introduce [GMT]'s Proposition 1.28 as Theorem 2.XX, and sketch its proof.

In Section 2 we provide a formal statement and proof of Theorem 2.XX.

INCREMENTSECTIONNUMBER(In Sections 2 through 5 we address the computer-related aspects of the proof. In Section 2, the method for describing the decomposition of the parameter space \mathcal{W} into sub-regions is given, and the conditions used to eliminate all but seven of the sub-regions are dis-

cussed. Near the end of this chapter, the first part of a detailed example is given. Eliminating a sub-region requires that a certain function is shown to be bounded appropriately over the entire sub-region. This is carried out by using a first-order Taylor approximation of the function together with a remainder bound. Our computer version of such a Taylor approximation with remainder bound is called an *AffApprox* and in Section 3, the relevant theory is developed. At this point, the detailed example of Section 2 can be completed.

In Sections 4 and 5, round-off error analysis appropriate to our set-up is introduced. Specifically, in Section 5, round-off error is incorporated into the *AffApprox* formulas introduced in Section 3. The proofs here require an analysis of round-off error for complex numbers, which is carried out in Section 4.)

In Section 7 we WORDSMITH(breadcrumbs about finding killerwords).

Finally, in Section 8, we present an updated version of the code used to check that the proof is valid, with self-contained copies of the proofs required for the reader check its own validity.

Acknowledgements. FIXME

1. Technical Introduction

DOES THIS BELONG IN BREADCRUMBS?

Here is a brief description of why Theorem 2.XX might be amenable to computer-assisted proof. If a shortest geodesic δ in a hyperbolic 3-manifold N does not have a $\ln(3)/2$ tube then there is a 2-generator subgroup G of $\pi_1(N) = \Gamma$ which also does not have that property. Specifically, take G generated by f and w , with $f \in \Gamma$ a primitive hyperbolic isometry whose fixed axis $\delta_0 \subset \mathbf{H}^3$ projects to δ , and with $w \in \Gamma$ a hyperbolic isometry which takes δ_0 to a nearest translate. Then, after identifying $N = \mathbf{H}^3/\Gamma$ and letting $Z = \mathbf{H}^3/G$, we see that the shortest geodesic in Z (which corresponds to δ) does not have a $\ln(3)/2$ tube. Thus, to understand solid tubes around shortest geodesics in hyperbolic 3-manifolds, we need to understand appropriate 2-generator groups, and this can be done by a parameter space analysis as follows. (Parameter space analyses are naturally amenable to computer proofs.)

The space of *relevant* (see Definition 1.12) 2-generator groups in $\text{Isom}_+(\mathbf{H}^3)$ is naturally parametrized by a subset \mathcal{P} of \mathbf{C}^3 . Each parameter corresponds to a 2-generator group G with specified generators f and w , and we call such a group a *marked group*. The marked groups of particular interest are those in which G is discrete, torsion-free, parabolic-free, f corresponds to a shortest geodesic δ , and w corresponds to a covering translation of a particular lift of δ to a nearest translate. We denote this set of particularly interesting marked groups by \mathcal{T} . We show that if $\text{tuberadius}(\delta) \leq \ln(3)/2$ in a hyperbolic

3-manifold N , then G must correspond to a parameter lying in one of seven small regions \mathcal{R}_n , $n = 0, \dots, 6$ in \mathcal{P} . With respect to this notation, we have:

PROPOSITION 1.28. $\mathcal{T} \cap (\mathcal{P} - \cup_{n=0,\dots,6} \mathcal{R}_n) = \emptyset$.

The full statement of Proposition 1.28 explicitly describes the seven small regions of the parameter space as well as some associated data.

Here is the idea of the proof. Roughly speaking, we subdivide \mathcal{P} into a billion small regions and show that all but the seven exceptional regions cannot contain a parameter corresponding to a “shortest/nearest” marked group. For example we would know that a region \mathcal{R} contained no such group if we knew that for each point $\rho \in \mathcal{R}$, $\text{Relength}(f_\rho) > \text{Relength}(w_\rho)$. (Here $\text{Relength}(f_\rho)$ (resp. $\text{Relength}(w_\rho)$) denotes the real translation length of the isometry of \mathbf{H}^3 corresponding to the element f (resp. w) in the marked group with parameter ρ .) This inequality would contradict the fact that f corresponds to δ which is a *shortest* geodesic. Similarly, there are *nearest* contradictions.

2. Killerwords and the parameter space

Notation and conventions 1.1. A hyperbolic 3-manifold is a Riemannian 3-manifold of constant sectional curvature -1 . All hyperbolic 3-manifolds under consideration will be closed and orientable. We will work in the upper-half-space model for hyperbolic 3-space: $\mathbf{H}^3 = \{(x, y, z) : z > 0\}$ with metric $ds_H = ds_E/z$. The distance between two points w and v in \mathbf{H}^3 will be denoted $\rho(w, v)$.

It is well known that $\text{Isom}_+(\mathbf{H}^3) = \text{PSL}(2, \mathbf{C})$, where an element of $\text{PSL}(2, \mathbf{C})$ acts as a Möbius transformation on the bounding (extended) complex plane and the extension to upper-half-space is the natural extension (see [Bea]). If M is a hyperbolic 3-manifold, then $M = \mathbf{H}^3/\Gamma$ where Γ is a discrete, torsion-free subgroup of $\text{PSL}(2, \mathbf{C})$.

For computational convenience, we will often normalize so that the (positive) z -axis is the axis of an isometry. As such, we set up some special notation. Let $B_{(0;\infty)}$ denote the oriented geodesic $\{(0, 0, z) : 0 < z < \infty\}$, with negative endpoint $(0, 0, 0)$. (An *endpoint* of an axis refers to a limit point of the axis on S_∞^2 .) Let $B_{(-1;1)}$ denote the oriented geodesic with negative endpoint $(-1, 0, 0)$ and positive endpoint $(1, 0, 0)$.

When working in a group G generated by f and w and looking at words in f, w, f^{-1}, w^{-1} we will often let F and W denote f^{-1} and w^{-1} , respectively.

Definition 2.2. If f is an isometry, then we define

$$\text{Relength}(f) = \inf\{\rho(w, f(w)) \mid w \in \mathbf{H}^3\}.$$

Thus $\text{Relength}(f) = 0$ if and only if f is either a parabolic or elliptic isometry. If $\text{Relength}(f) > 0$, then f is hyperbolic and maps a unique geodesic σ in \mathbf{H}^3

to itself. In that case σ is oriented (the negative end being the repelling fixed point on S_∞^2) and the isometry f is the composition of a rotation of $t \pmod{2\pi}$ radians along σ (the sign of the angle of rotation is determined by the right-hand rule) followed by a pure translation of \mathbf{H}^3 along σ of $l = \text{Relength}(f)$. We define $\text{length}(f) = l + it$, and call $A_f = \sigma$ the *axis* of f . Now, A_f is an oriented interval with endpoints in S_∞^2 , the orientation being induced from σ .

If the geodesic σ is given a fixed orientation, we define an $l + it$ translation f along σ to be a distance l translation in the positive direction, followed by a rotation of σ by t radians. Of course if $l < 0$, then each point of σ gets moved $-l$ in the negative direction. Also, via the right-hand rule, the orientation determines what is meant by a t -radian rotation. Thus if $l > 0$, the orientation induced on σ by f (as in the previous paragraph) equals the given orientation. If $l < 0$, then the induced orientation is opposite to the given orientation and f is a $-(l + it)$ translation of $-\sigma$ in the sense of the previous paragraph.

If f is elliptic, then f is a rotation of t radians where $0 \leq t \leq \pi$ about some oriented geodesic, and we define $\text{length}(f) = ti$. If f is parabolic or the identity, we define $\text{length}(f) = 0 + i0$. So, for all isometries we have that $\text{Relength} = \text{Re}(\text{length})$.

Definition 2.3. If G is a subgroup of $\text{Isom}_+(\mathbf{H}^3)$, then we say that f is a *shortest* element in G if $f \neq \text{id}$ and $\text{Relength}(f) \leq \text{Relength}(g)$ for all $g \in G$, $g \neq \text{id}$.

Definition 2.4. If σ, τ are disjoint oriented geodesics in \mathbf{H}^3 which do not meet at infinity, then define $\text{distance}(\sigma, \tau) = \text{length}(w)$ where $w \in \text{Isom}_+(\mathbf{H}^3)$ is the hyperbolic element which translates \mathbf{H}^3 along the unique common perpendicular between σ and τ and which takes the oriented geodesic σ to the oriented geodesic τ . The oriented common perpendicular from σ to τ is called the *orthocurve* between σ and τ . The *ortholine* between σ and τ is the complete oriented geodesic in \mathbf{H}^3 which contains the orthocurve between σ and τ .

If σ and τ intersect at one point in \mathbf{H}^3 then there is an elliptic isometry w taking σ to τ fixing $\sigma \cap \tau$. Again, define $\text{distance}(\sigma, \tau) = \text{length}(w)$. In this case, the orthocurve is the point $\sigma \cap \tau$, and the ortholine O from σ to τ is oriented so that σ, τ, O form a right-handed frame.

If σ and τ intersect at infinity, then there is no unique common perpendicular, hence no ortholine, and we define $\text{distance}(\sigma, \tau) = 0 + i0$, or $0 + i\pi$ depending on whether or not σ and τ point in the same direction at their intersection point(s) at infinity.

Define $\text{Redistance} = \text{Re}(\text{distance})$.

As defined, Redistance is nonnegative. In Definition 1.8 and in Sections 2 and 3, it will be useful to have a broader definition. Given an oriented geodesic α in \mathbf{H}^3 orthogonal to oriented geodesics β and γ , define $d_\alpha(\beta, \gamma) \in \mathbf{C}$ where a $d_\alpha(\beta, \gamma)$ translation of \mathbf{H}^3 along α takes β to γ .

Definition 2.5. A tube of radius r about a geodesic δ_0 in \mathbf{H}^3 is $\{w \in \mathbf{H}^3 \mid \rho(w, v) \leq r \text{ for some } v \in \delta_0\}$. If δ is a simple closed geodesic in the hyperbolic 3-manifold N and if $\{\delta_i\}$ is the set of pre-images of δ in \mathbf{H}^3 , then define $\text{tuberadius}(\delta) = \frac{1}{2} \min\{\text{Redistance}(\delta_i, \delta_j) \mid i \neq j\}$. If $r = \text{tuberadius}(\delta)$, then define a *maximal tube* about δ to be the image of a tube of radius r about δ_0 . Note that $\text{tuberadius}(\delta) = \sup\{r \mid \text{there exists an embedded tubular neighborhood of radius } r \text{ about } \delta\}$.

Definition 2.6. Our desire to understand tuberadii about closed geodesics, and especially about a simple closed geodesic δ , leads us to investigate certain 2-generator subgroups $G = \langle f, w \rangle$ of $\text{Isom}_+(\mathbf{H}^3)$ with the generator f corresponding to a primitive isometry fixing δ_0 and the generator w corresponding to an element taking δ_0 to its nearest covering translate. We investigate these 2-generator groups by using certain subsets of \mathbf{C}^3 as parameter spaces.

A *marked (2-generator) group* is a triple $\{G, f, w\}$ consisting of a 2-generator subgroup G of $\text{Isom}_+(\mathbf{H}^3)$ and an ordered pair of isometries f, w of \mathbf{H}^3 which generate G such that $\text{Relength}(f) > 0$ and if A_f is the axis of f , then $w(A_f) \cap A_f = \emptyset$ (here, intersection is taken in $\mathbf{H}^3 \cup S_\infty^2$). Two marked groups $\{G_1, f_1, w_1\}$ and $\{G_2, f_2, w_2\}$ are *conjugate* if G_1 and G_2 are conjugate via an element of $\text{Isom}_+(\mathbf{H}^3)$ and this conjugating element takes f_1 to f_2 and w_1 to w_2 . Within any conjugacy class of marked groups is a unique

normalized element $\{G, f, w\}$ where f is a positive translation along the (oriented) geodesic $B_{(0;\infty)}$, and the orthocurve from $w^{-1}(B_{(0;\infty)})$ to $B_{(0;\infty)}$ lies on $B_{(-1;1)}$ on the negative side of $B_{(-1;1)} \cap B_{(0;\infty)}$. To minimize notation, we will frequently equate a conjugacy class with its normal representative.

Given $(L, D, R) = (l + it, d + ib, r + ia) \in \mathbf{C}^3$ with $l > 0$, $d > 0$, one can associate a group G generated by elements f and w as follows. Define f to be an $l + it$ translation along $B_{(0;\infty)}$ and w to be a $d + ib$ translation along $B_{(-1;1)}$ followed by an $r + ia$ translation along $B_{(0;\infty)}$ (here, r can be negative, in which case this is equivalent to a $-r - ia$ translation along $-B_{(0;\infty)}$). Conversely if $\{G, f, w\}$ is a normalized marked group then f is an L translation of $B_{(0;\infty)}$ and w is a D translation of $B_{(-1;1)}$ followed by an R translation of $B_{(0;\infty)}$. Thus $\mathcal{P}' = \{(l + it, d + ib, r + ia) \in \mathbf{C}^3 \mid l > 0, d > 0\}$ parametrizes the set of conjugacy classes of marked groups. In particular, the parametrization is surjective and locally one-to-one.

We are primarily interested in the set $\mathcal{T}' \subset \mathcal{P}'$ which parametrizes all conjugacy classes of marked groups $\{G, f, w\}$ for which f is a shortest element of G which (positively) translates $B_{(0;\infty)}$ and $w \in G$ takes $B_{(0;\infty)}$ to a *nearest* translate $w(B_{(0;\infty)})$ such that $-\text{Relength}(f)/2 < \text{Re}(d_{B_{(0;\infty)}})$ (ortholine from $w^{-1}(B_{(0;\infty)})$ to $B_{(0;\infty)}$), (ortholine from $B_{(0;\infty)}$ to $w(B_{(0;\infty)})$)) $\leq \text{Relength}(f)/2$. See Figure 1.1. Note that because f is shortest and $\text{Relength}(f) > 0$, it follows that G must be discrete, torsion-free, and parabolic-free.

Remark 2.7. \mathcal{T}' consists of those parameters corresponding to marked groups $\{G, f, w\}$ such that l is the real length of a shortest element of G , d is the real distance between $B_{(0;\infty)}$ and a nearest translate, and $-l/2 < r \leq l/2$. In what follows, it is essential to remember that an element α of \mathcal{P}' corresponds not only to a group G , but to a marked group. To further establish the point, we note that, for elements of \mathcal{T}' , the parameter l is an invariant of G alone (that is, l is the shortest real length of an element of G), while the parameter d is determined by G and f (that is, the notion of “nearest” used to define w in the definition of \mathcal{T}' requires a choice of f).

As mentioned in the introduction to this paper, we are only interested in the subset of \mathcal{T}' corresponding to parameters α with $d \leq \ln(3)$. The following two propositions imply this subset of \mathcal{T}' lives in a compact subset of \mathcal{P}' .

Figure 1.1

Definition 2.8. Let $\mathcal{P} \subset \mathcal{P}'$ be the set of those parameters $\alpha = (l + it, d + ib, r + ia)$ such that

- a) $0.0978 \leq l \leq 1.289785$,
- b) $l/2 \leq d \leq \ln(3)$,
- c) $0 \leq r \leq l/2$,
- d) $-\pi \leq t \leq 0$,
- e) $-\pi \leq b \leq \pi$,
- f) $-\pi \leq a \leq \pi$.

Define $\mathcal{T} = \mathcal{T}' \cap \mathcal{P}$.

The point of the definition of \mathcal{P} and \mathcal{T} is as follows. We want to analyze by computer the relationship between lengths of shortest geodesics and their tuberadii in hyperbolic 3-manifolds. We were naturally led to the parameter space \mathcal{P}' and its subset \mathcal{T}' . But \mathcal{P}' is problematic from the computational view-point because it is noncompact. We wish to replace \mathcal{P}' by \mathcal{P} which is compact, and \mathcal{T}' by \mathcal{T} in our computer analysis. This is carried out in Lemma 1.13. Note that we worked to make \mathcal{P} as small as reasonable to save computation time; for example, the t and r restrictions above cut down the parameter space by a factor of 4 over the obvious t and r restrictions.

By [G; Lemma 5.9] (or see Example A.3 in the appendix) a closed orientable hyperbolic 3-manifold N satisfies the insulator condition provided that $\text{tuberadius}(\delta) > \ln(3)/2$ for some closed geodesic $\delta \subset N$. Thus we are led to consider:

Definition 2.9. A word h in w, f, w^{-1}, f^{-1} for which statement a) (resp. b)) in Remark 1.17 holds for each $\beta \in \mathcal{P}_i$ and for which h_β is not a power of f_β for each $\beta \in \mathcal{P}_i$ is called a *killerword* for \mathcal{P}_i with respect to contradiction a) (resp. b)).

Summary 2.10. With seven exceptions, to each of the approximately one billion regions partitioning \mathcal{P} , we will associate a killerword and a contradiction.

Remark 2.11. Computers are well suited for partitioning a set such as \mathcal{P} into many regions $\{\mathcal{P}_i\}$, and finding a killerword h_i which eliminates all $\alpha_i \in \mathcal{P}_i$ due to contradiction C_i . Depending on the contradiction, we find computable expressions for approximations of the values of $\text{Relength}(h_\beta)$ or $\text{Redistance}(h_\beta(B_{(0;\infty)}), B_{(0;\infty)})$ and thus use the computer to eliminate all of \mathcal{P}_i .

Definition 1.22. Let

$$\mathcal{W} = \{(x_0, x_1, x_2, x_3, x_4, x_5) : |x_i| \leq 4 \times 2^{(5-i)/6} \text{ for } i = 0, 1, 2, 3, 4, 5\}$$

$$\supset \exp(\mathcal{P}) = \{(x_0, x_1, x_2, x_3, x_4, x_5) \mid x_0 + ix_3 = \exp(e), x_1 + ix_4 = \exp(f), \\ x_2 + ix_5 = \exp(g) \text{ where } (e, f, g) \in \mathcal{P}\}$$

and let

$$\mathcal{S} = \exp(\mathcal{T}).$$

As we are taking \exp of the various complex co-ordinates, it is notationally convenient to replace our complex parameters $L = l + it$, $D = d + ib$, $R = r + ia$ by exponentiated versions. That is, let

$$L' = \exp(L) = \exp(l + it), \quad D' = \exp(D) = \exp(d + ib), \\ R' = \exp(R) = \exp(r + ia).$$

Remarks 1.23. i) We work with \mathcal{W} instead of $\exp(\mathcal{P})$ because we want our initial parameter space to be a (6-dimensional) box that is easily subdivided. This has the side effect that certain regions (sub-boxes) \mathcal{W}_i of \mathcal{W} will be eliminated because they are outside of $\exp(\mathcal{P})$ not because of the analogues of conditions a) and b) in Remark 1.17. The entire collection of conditions is given in Chapter 5.

ii) The presence of the factor $2^{(5-i)/6}$ in the definition of \mathcal{W} is explained in Construction 5.3. Briefly, the main reason for including it is to make the shape of regions stay as uniform and “round” as possible under subdivision. This makes the Taylor approximations efficient, hence fast.

iii) We chose the co-ordinates of \mathcal{W} so that $L' = x_0 + ix_3$, $D' = x_1 + ix_4$, $R' = x_2 + ix_5$ to gain a mild computer advantage.

LEMMA 2.15. *If $(L', D', R') \in \mathcal{W}$ and $\{G, f, w\}$ is the associated normalized marked group, then f and w have matrix representatives*

$$\text{a) } f = \begin{pmatrix} \sqrt{L'} & 0 \\ 0 & 1/\sqrt{L'} \end{pmatrix},$$

$$\text{b) } w = \begin{pmatrix} \sqrt{R'} * ch & \sqrt{R'} * sh \\ sh/\sqrt{R'} & ch/\sqrt{R'} \end{pmatrix}$$

where $ch = (\sqrt{D'} + 1/\sqrt{D'})/2$ and $sh = (\sqrt{D'} - 1/\sqrt{D'})/2$.

Proof. a) In our set-up the (oriented) axis of f is $B_{(0;\infty)}$. As such, f corresponds to a diagonal matrix, with diagonal entries p and p^{-1} , with $|p| > 1$. The action of f on the bounding complex plane is simply multiplication by p^2 . Extending this action to upper-half-space in the natural way rotates the z -axis by angle $\arg(p^2)$ and sends $(0, 0, 1)$ to $(0, 0, |p|^2)$. Thus,

$$\text{Im}(\text{length}(f)) = \arg(p^2) = \text{Im}(\ln(p^2))$$

and, using the hyperbolic metric,

$$\text{Re}(\text{length}(f)) = \ln(|p|^2) = \text{Re}(\ln(p^2)).$$

That is, $\text{length}(f) = \ln(p^2)$ and

$$p = \pm \exp(\text{length}(f)/2) = \pm \sqrt{\exp(\text{length}(f))} = \pm \sqrt{\exp(L)} = \pm \sqrt{L'}.$$

Now, we take the positive square root (taking the negative square root produces the other lift from $\text{PSL}(2, \mathbf{C})$ to $\text{SL}(2, \mathbf{C})$).

b) $w = \beta \circ \alpha$ where β is translation of distance R along $B_{(0;\infty)}$ and α is translation of distance D along $B_{(-1;1)}$. Thus, a matrix representative of β is

$$\begin{pmatrix} \sqrt{R'} & 0 \\ 0 & 1/\sqrt{R'} \end{pmatrix}$$

and a matrix representative of α can be computed to be

$$\begin{pmatrix} \cosh(D/2) & \sinh(D/2) \\ \sinh(D/2) & \cosh(D/2) \end{pmatrix}.$$

But $\cosh(D/2) = (\exp(D/2) + \exp(-D/2))/2 = (\sqrt{D'} + 1/\sqrt{D'})/2 = ch$ and

similarly for sh . Thus,

$$\alpha = \begin{pmatrix} ch & sh \\ sh & ch \end{pmatrix}$$

and b) follows by matrix multiplication. \square

LEMMA 2.16. *If $h \in \text{Isom}_+(\mathbf{H}^3)$ is represented by the matrix*

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \in \text{SL}(2, \mathbf{C}),$$

then

- a) $\exp(\text{Relength}(h)) = |\text{trace}(A)/2 \pm \sqrt{(\text{trace}(A)/2)^2 - 1}|^2$,
- b) $\exp(\text{Redistance}(h(B_{(0;\infty)}), B_{(0;\infty)})) = |\text{orthotrace}(A) \pm \sqrt{(\text{orthotrace}(A))^2 - 1}|$ where $\text{orthotrace}(A) = ad + bc$.

In both cases, the $+$, $-$ produce reciprocal values for the right-hand side, and we take the one producing the larger value, unless the value is 1, in which case there is no need to choose.

Proof. a) If A is elliptic or parabolic, the proof is straightforward (the trace of a parabolic is ± 2 while the trace of an elliptic is a real number between 2 and -2).

We assume A is hyperbolic. Because trace is a conjugacy invariant, we can assume the oriented axis of A is $B_{(0;\infty)}$. Thus A is a diagonal matrix with p and p^{-1} along the diagonal with $|p| > 1$, and, as in the proof of Lemma 1.24, we see that $\exp(\text{length}(h)) = p^2$. Of course, $\text{trace}(A) = p + p^{-1}$, and it is easy enough to solve for p . Specifically, $p = \text{trace}(A)/2 \pm \sqrt{(\text{trace}(A)/2)^2 - 1}$. Thus,

$$\begin{aligned} \exp(\text{Relength}(h)) &= |\exp(\text{length}(h))| = |p|^2 \\ &= |(\text{trace}(A)/2) \pm \sqrt{(\text{trace}(A)/2)^2 - 1}|^2. \end{aligned}$$

b) If $B_{(0;\infty)}$ and $h(B_{(0;\infty)})$ intersect at infinity, then the proof is straightforward. For example, if h fixes the point $(0, 0, 0)$ at infinity, then $c = 0$, $ad = 1$ and the formula holds. Similarly for the other cases in which $B_{(0;\infty)}$ and $h(B_{(0;\infty)})$ intersect at infinity.

We assume $B_{(0;\infty)}$ and $h(B_{(0;\infty)})$ do not intersect at infinity. We will compute the length of k , the square of the transformation taking $B_{(0;\infty)}$ to $h(B_{(0;\infty)})$ along their ortholine. Let τ be 180-degree rotation about $B_{(0;\infty)}$, then $(h \circ \tau \circ h^{-1})$ is 180-degree rotation about $h(B_{(0;\infty)})$, and we have that $k = (h \circ \tau \circ h^{-1}) \circ \tau$. Now, τ and h are represented by the matrices

$$\begin{pmatrix} i & 0 \\ 0 & -i \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} a & b \\ c & d \end{pmatrix} \in \text{SL}(2, \mathbf{C}).$$

Hence, $k = (h \circ \tau \circ h^{-1}) \circ \tau$ can be computed to have matrix representation

$$\begin{pmatrix} ad + bc & 2ab \\ 2cd & ad + bc \end{pmatrix}.$$

Thus,

$$\begin{aligned} & \exp(\text{Redistance}(h(B_{(0;\infty)}), B_{(0;\infty)})) \\ &= \exp(\text{Relength}(k)/2) = \sqrt{|\exp(\text{length}(k))|} \\ &= |(\text{trace}(k)/2) \pm \sqrt{(\text{trace}(k)/2)^2 - 1}| \\ &= |(ad + bc) \pm \sqrt{(ad + bc)^2 - 1}|. \end{aligned}$$

□

Remarks 2.17. i) It follows from Lemma 1.25 that if h is a word in f, w, f^{-1}, w^{-1} , then for any parameter value $\alpha \in \mathcal{W}$,

$$\exp(\text{Relength}(h_\alpha)), \text{ and } \exp(\text{Redistance}(h_\alpha(B_{(0;\infty)}), B_{(0;\infty)}))$$

can be computed using only the operations $+, -, \times, /$, $\sqrt{}$.

ii) During the course of the computer work needed to prove the main theorems, the parameter space \mathcal{W} was decomposed into sub-boxes by computer via a recursive subdivision process: Given a sub-box being analyzed, either it can be *killed directly* (that is, eliminated by a killerword and associated condition as described in Remark 1.17, or for the trivial reason described in Remark 1.23i), or it cannot. If it cannot be killed directly, it is subdivided in half by a hyperplane $\{x_i = c\}$ (where i runs through the various co-ordinate dimensions cyclically) and the two pieces are analyzed separately, and so on.

As such, a sub-box of \mathcal{W} can be described by a sequence of 0's and 1's where 0 means “take the lesser x_i values” and 1 means “take the greater x_i values.” For the decomposition of \mathcal{W} into sub-boxes, all the sub-box descriptions could be neatly encoded into one tree (although in practice we found it preferable to use several trees to describe the entire decomposition. See §5).

iii) In the following proposition, seven *exceptional boxes* are described as sequences of 0's and 1's. Four of the exceptional boxes— X_0, X_4, X_5, X_6 —are each the union of two abutting sub-boxes, $X_0 = X_{0a} \cup X_{0b}$ and so on. It is a pleasant exercise to work through the fact that they abut. It should be noted that had the set-up for \mathcal{W} been different, more sub-boxes (or perhaps fewer) might have been needed to construct the seven exceptional regions.

It is also a pleasant exercise to calculate by hand the co-ordinate ranges of the various sub-boxes. For example, the range of the last co-ordinate (i.e., x_5) of the sub-box

$$X_{6a} = 111000000001000111 \ 111111110101001111 \ 011111010111111111 \\ 1100010010110001111 \ 0$$

is found by taking the 6th entry, the 12th entry, the 18th entry, and so on. These entries are 011111111111. The first entry (0) means take the lesser x_5 values, and produces the interval $[-4, 0]$. The second entry (1) means take the greater x_5 values, and produces the interval $[-2, 0]$. The third entry (1) produces $[-1, 0]$. Continuing, we see that X_{6a} has $-2^{-9} \leq x_5 = \text{Im}(R') \leq 0$. The other co-ordinates can be computed in the same fashion, although they must at the end be multiplied by the factor $2^{(5-i)/6}$ (see the definition of the initial box \mathcal{W}). The range of co-ordinate values for each exceptional box X_0, X_1, \dots, X_6 is given in Table 1.1 (a limited number of significant digits is given), and then a range of co-ordinates for exceptional regions (in \mathcal{P}) $\mathcal{R}_i \supset \exp^{-1}(X_i)$ is given (see Remarks 1.30i) and 1.30ii) in Table 1.2. (Note that this use of the symbol \mathcal{R}_i differs slightly from the use in §0.) Finally, two *quasi-relators* are given in Proposition 1.28 for each exceptional box X_0, X_1, \dots, X_6 (see the next definition).

Definition 2.18. A *quasi-relator* in a sub-box X of \mathcal{W} is a word in $f, w, F = f^{-1}, W = w^{-1}$ that is close to the identity throughout X and experimentally appears to be converging to the identity at some point in X . In particular, a quasi-relator rigorously has Relength less than that of f at all points in X .

PROPOSITION 2.19. *Within the parameter space \mathcal{W} but outside the seven exceptional boxes there are no parameter points corresponding to marked groups $\{G, f, w\}$ where G is discrete, torsion-free and parabolic-free; f corresponds to a shortest geodesic δ of tuberradius $\leq \ln(3)/2$; and w takes a particular lift of δ to a nearest translate. Specifically, $\mathcal{S} \cap (\mathcal{W} - \bigcup_{n=0, \dots, 6} X_n) = \emptyset$ where the X_n are the exceptional boxes*

$$X_0 = X_{0a} \cup X_{0b},$$

$$X_{0a} = 001000110111110001 \ 101001010101011001 \ 011011010111101101 \\ 100001101101000111 \ 010001110101100101 \ 1101110111110100,$$

$$X_{0b} = 001001110110110000 \ 101000010100011000 \ 011010010110101100 \\ 100000101100000110 \ 010000110100100100 \ 1101100111100100,$$

X_0 *quasi-relators:*

$$r_1 = fwFwwFwfwfww, \\ r_2 = FwfwfWfwfw,$$

$$X_1 = 001000110001110110 \ 011101000110111110 \ 100010110000100011 \\ 101101001101001000 \ 110101011000000100 \ 000.$$

X_1 *quasi-relators:*

$$\begin{aligned} r_1 &= FFwFWFWfWFWFwFFww, \\ r_2 &= FFwwFwfWfWfWfWfWfW, \end{aligned}$$

$$\begin{aligned} X_2 = & 001000110101010010 \ 101010110001100101 \ 110111100001101010 \\ & 111100100000010001 \ 111100, \end{aligned}$$

X_2 *quasi-relators:*

$$\begin{aligned} r_1 &= FwfwfWfWfWfWfWfW, \\ r_2 &= FFwFFwFwFwFwFwFwFw, \end{aligned}$$

$$\begin{aligned} X_3 = & 111000000001000110 \ 011011101101011000 \ 111101011110001100 \\ & 111111100110110000 \ 0000100010100010, \end{aligned}$$

X_3 *quasi-relators:*

$$\begin{aligned} r_1 &= FFwfwFFwwFWFwFWfWFWfWFWfWFWfWFWfW, \\ r_2 &= FFwfwFwfwfWfWfWfWfWfWfWfWfWFFwwFWFwFw, \end{aligned}$$

$$X_4 = X_{4a} \cup X_{4b},$$

$$\begin{aligned} X_{4a} = & 111000000001000110 \ 011001001111101010 \ 011110110110111101 \\ & 10001111110110110 \ 100001111101, \end{aligned}$$

$$\begin{aligned} X_{4b} = & 111000000001000110 \ 011001001111101010 \ 111110010110011101 \\ & 000011011110010110 \ 000001011101, \end{aligned}$$

X_4 *quasi-relators:*

$$\begin{aligned} r_1 &= FFwfwFwfwfWfWfWfWfWfWFFwwFWFwFWFwFw, \\ r_2 &= FFwfwFwfwfWFFwwFWFwFWfWFWfWFWfWFWFwFw, \end{aligned}$$

$$X_5 = X_{5a} \cup X_{5b},$$

$$\begin{aligned} X_{5a} = & 001000110001110111 \ 001111000101111111 \ 101111100111001111 \\ & 000001111011110111 \ 1, \end{aligned}$$

$$\begin{aligned} X_{5b} = & 001001110000110110 \ 001110000100111110 \ 101110100110001110 \\ & 000000111010110110 \ 1, \end{aligned}$$

X_5 *quasi-relators:*

$$\begin{aligned} r_1 &= FwFWFwFwfwfWfWfW, \\ r_2 &= FwfwfWfWFWfWfWfW, \end{aligned}$$

$$X_6 = X_{6a} \cup X_{6b},$$

$$\begin{aligned} X_{6a} = & 111000000001000111 \ 111111110101001111 \ 011111010111111111 \\ & 110001001011000111 \ 0, \end{aligned}$$

$$\begin{aligned} X_{6b} = & 111001000000000110 \ 111110110100001110 \ 011110010110111110 \\ & 110000001010000110 \ 0, \end{aligned}$$

X_6 *quasi-relators:*

$$\begin{aligned} r_1 &= FWFwFWfWFWFWfWfw, \\ r_2 &= FWFwfwFwfwfWfWfWfW. \end{aligned}$$

Proof. The proof follows along the lines presented in Remark 1.17. Two computer files contain the data needed for the proof. The first computer file describes the partition of \mathcal{W} into sub-boxes and attaches an integer to each such sub-box, and the second file, called “conditionlist” is an ordered list of conditions and killerwords. The integer associated to a sub-box in the first file describes the numbered condition/killerword from conditionlist that will eliminate the sub-box in question (other than those corresponding to the X_i). A computer program named *verify* shows that the conditions and killerwords in question actually do kill off their associated sub-boxes (see Section 5 for more details). This computer program addresses the issues of Remark 1.20. The code for *verify* is available at the *Annals* web site.

In addition, a mild modification of *verify* showed that the listed words were quasi-relators for the given sub-boxes. \square

3. Conditions and sub-boxes

In this section we expand on some topics mentioned briefly in Section 1. As such, it would be useful to look again at FIXME(Definitions 1.8, 1.12, and 1.22, and Remark 1.9), where \mathcal{P} , \mathcal{T} and their partners (under exponentiation) \mathcal{W} , \mathcal{S} are introduced, and at FIXME(Definition 1.18) where the notion of a killerword is introduced (the definition is phrased in terms of \mathcal{P} , but the definition also makes sense for \mathcal{W}). Note that working with the region \mathcal{P} is intuitively appealing, but working with the box \mathcal{W} is vastly superior computationally (see Remark 1.21).

FIXME(Theorem 0.2), hence FIXME(Theorem 0.1), follows from FIXME(Propositions 1.28, 2.8, 3.1, and 3.2). The computational aspects of the proofs of each of these propositions are similar. We will focus on FIXME(Proposition 1.28) here. Its proof amounts to decomposing \mathcal{W} into a collection of sub-boxes of two types:

- 1) The 11 sub-boxes which comprise the exceptional boxes X_0, X_1, \dots, X_6 .
- 2) Sub-boxes each of which has an associated *condition* that will describe how to kill that entire sub-box, perhaps with the help of a killerword. To *kill* a sub-box means to show that $\mathcal{S} - \bigcup_{n=0, \dots, 6} X_n$ has no point in the sub-box.

The set-up for efficiently describing these sub-boxes will be given in Construction 5.3.

We now list the conditions used to kill the nonexceptional sub-boxes. There are two types of conditions: the trivial and the interesting. The trivial conditions kill sub-boxes in \mathcal{W} since the sub-boxes in question miss $\exp(\mathcal{P})$.

The interesting conditions are where the real work is done, and they require a killerword in f, w, f^{-1}, w^{-1} to work their magic (see FIXME(Remark 1.17)).

To be consistent with the computer program *verify* we use the following notation: $L' = z_0 + iz_3$, $D' = z_1 + iz_4$, and $R' = z_2 + iz_5$. Here $(L', D', R') \in \mathcal{W}$ and $L' = \exp(L) = \exp(l + it)$, $D' = \exp(D) = \exp(d + ib)$, $R' = \exp(R) = \exp(r + ia)$.

The trivial conditions FIXME(5.1) .

Condition ‘s’ (short): Tests that all points in the sub-box have $|z_0 + iz_3| < 1.10274$. This ensures that

$$\exp(l) = |\exp(L)| = |L'| = |z_0 + iz_3| < 1.10274 < \exp(0.0978),$$

and FIXME(Definition 1.12) tells us that we are outside of $\exp(\mathcal{P})$.

Condition ‘l’ (long): Tests that all points in the sub-box have $|z_0 + iz_3| > 3.63201$. This ensures that

$$\exp(l) = |\exp(L)| = |L'| = |z_0 + iz_3| > 3.63201 > \exp(1.289785)$$

and we are outside of $\exp(\mathcal{P})$.

Condition ‘n’ (near): Tests that all points in the sub-box have $|z_1 + iz_4| < 1$. This ensures that

$$\exp(d) = |\exp(D)| = |D'| = |z_1 + iz_4| < 1 = \exp(0)$$

and we are outside of $\exp(\mathcal{P})$.

Condition ‘f’ (far): Tests that all points in the sub-box have $|z_1 + iz_4| > 3$. This ensures that

$$\exp(d) = |\exp(D)| = |D'| = |z_1 + iz_4| > 3 = \exp(\ln 3)$$

and we are outside of $\exp(\mathcal{P})$.

Condition ‘w’ (whirle big): Tests that all points in the sub-box have $|z_2 + iz_5|^2 > |z_0 + iz_3|$. This ensures that

$$\exp(r) = |\exp(R)| = |R'| = |z_2 + iz_5| > \sqrt{|z_0 + iz_3|} = \sqrt{\exp(l)} = \exp(l/2)$$

and we are outside of $\exp(\mathcal{P})$.

Condition ‘W’ (whirle small): Tests that all points in the sub-box have $|z_2 + iz_5| < 1$. This ensures that

$$\exp(r) = |\exp(R)| = |R'| = |z_2 + iz_5| < 1 = \exp(0)$$

and we are outside of $\exp(\mathcal{P})$.

The interesting conditions **FIXME(5.2)** .

Condition ‘L’: This condition comes equipped with a killerword k in f, w, f^{-1}, w^{-1} , and tests that all points in the sub-box have $|\exp(\text{length}(k))| < |L'| = |\exp(L)|$, where $\text{length}(k)$ means the length of the isometry determined by k . This, of course, contradicts the fact that L is the length of the shortest geodesic.

It is easy to carry out the test $|\exp(\text{length}(k))| < |L'|$ because Lemma 1.25a) can be used. Note that in *verify* the function which computes $\exp(\text{length})$ is called *length*.

Of course, Condition ‘L’ also checks that the isometry corresponding to the word k is not the identity.

Condition ‘O’: This condition comes equipped with a killerword k in f, w, f^{-1}, w^{-1} , and tests that all points in the sub-box have

$$|\exp(\text{distance}(k(B_{(0,\infty)}), B_{(0,\infty)}))| < |D'| = |\exp(D)|.$$

(Recall that $B_{(0,\infty)}$ denotes the oriented geodesic $\{(0,0,z) : 0 < z < \infty\}$, with negative endpoint $(0,0,0)$.) This, of course, contradicts the “nearest” condition.

It is easy to carry out the test $|\exp(\text{distance}(k(B_{(0,\infty)}), B_{(0,\infty)}))| < |D'|$ because Lemma 1.25b) can be used. Note that in *verify* the function which computes the quantity $\exp(\text{distance}(k(B_{(0,\infty)}), B_{(0,\infty)}))$ is called *orthodist*.

Also, Condition ‘O’ checks that the isometry corresponding to the word k does not take the axis of f to itself.

Condition ‘2’: This is just the ‘L’ condition without the “not-the-identity” check, but with the additional proviso that the killerword k is of the form $f^p w^q$. This ensures that k is not the identity, because for k to be the identity f and w would have to have the same axis, which contradicts the fact that d can be taken to be greater than or equal to $l/4$.

Condition ‘conjugate’: There is one other condition that is used to eliminate points in \mathcal{W} . Following FIXME(Definition 1.12) (and FIXME(Lemma 1.13)) we eliminate all boxes with $0 < t \leq \pi$. Of course, after exponentiating $L = l + it$, this corresponds to eliminating all boxes with $z_3 > 0$. Specifically, we toss all sub-boxes of \mathcal{W} whose fourth entry is a 1. This condition does not appear in *verify* because it is applied “outside” of these programs, as described in Construction FIXME(5.3).

Construction 5.3: We now give the method for describing the roughly 930 million sub-boxes that the initial box \mathcal{W} is subdivided into.

All sub-boxes are obtained by subdivision of a previous sub-box along a real hyper-plane midway between parallel faces of the sub-box before subdivision. Of course, these midway planes are of the form $x_i = \text{a constant}$. We use 0’s and 1’s to describe which half of a subdivided sub-box to take (0 corresponds to lesser x_i values). For example, 0 describes the sub-box

$$\mathcal{W} \cap \{(x_0, x_1, x_2, x_3, x_4, x_5) : x_0 \leq 0\},$$

010 describes the sub-box

$$W \cap \{(x_0, x_1, x_2, x_3, x_4, x_5) : x_0 \leq 0, x_1 \geq 0, x_2 \leq 0\},$$

and so on.

In this way, we get a one-to-one correspondence between strings and sub-boxes. If s is a string of 0’s and 1’s, then let $Z(s)$ denote the sub-box corresponding to s . The range of values for the i^{th} coordinate in the sub-box $Z(s)$ is related to the binary fraction $0.s_i s_{i+6} \dots s_{i+6k}$. The two sub-boxes gotten from subdividing $Z(s)$ are $Z(s0)$ and $Z(s1)$.

The directions of subdivision cycle among the various coordinate axes: the n^{th} subdivision is across the $(n \bmod 6)^{\text{th}}$ axis. The dimensions of the top-level box \mathcal{W} were chosen so that subdivision is always done across the longest dimension of the box, and so that all of the sub-boxes are similar. The dimensions of \mathcal{W} have the beneficial effect of making the sub-boxes as “round” as possible, hence making the Taylor approximation calculations efficient and fast. This explains the factor of $2^{(5-i)/6}$ in Definition FIXME(1.22) .

To kill a sub-box $Z(s)$, the checker program has two (recursive) options: use a condition and, if necessary, an associated killerword to kill $Z(s)$ directly, or first kill $Z(s0)$ and then kill $Z(s1)$. At this point, it may seem as if the second option is not necessary, because surely a condition which kills two halves also kills the whole. The answer to this has been hinted at in FIXME(Remarks 1.17 and 1.35) where it is noted that our evaluation of a function arising from a killerword is via first-order Taylor approximation, complete with remainder/error term. (Note that the remainder/error term incorporates bounds on both the theoretical error arising from using a first-order Taylor approximation to approximate a function, and the accumulated round-off error; see

FIXME(Sections 6, 7, and 8) .) Even if a killerword could theoretically kill off a sub-box, it is quite possible that our first-order Taylor approximation approach would not be able to prove this because its remainder/error term is too large. However, if we subdivide the sub-box, then the first-order Taylor approximations on the two halves should be more accurate. Thus, we want to have the recursive subdivision option at our disposal. Note that because the checker program does in fact do such recursive subdivisions, the actual number of sub-boxes in the ultimate subdivision is larger (perhaps substantially larger) than the 930 million sub-boxes of the initial data tree.

It is also possible that the checker program will employ neither of the two options described in the previous paragraph, and will instead employ a third option: do not kill $Z(s)$, and instead mark s as omitted. Any omitted sub-boxes are checked with another instance of the checker program, unless the sub-box is one of the 11 exceptional sub-boxes (which produce the seven exceptional boxes after joining abutters). Note that according to the definition of “kill” given at the beginning of this section, the exceptional boxes are automatically killed.

Thus, a typical output from *verify* would be

verified000000111101111111
 $-\{0000001111011111110 \ 000000111101111111110\}.$

which means that the sub-box $Z(000000111101111111)$ was killed except for its sub-boxes $Z(0000001111011111110)$ and $Z(000000111101111111110)$. The output

verified00000011110111111110 $-\{ \}$.

and

verified000000111101111111110 $-\{ \}$.

shows that these sub-boxes were subsequently killed as well, and thus the entire sub-box $Z(000000111101111111)$ has been killed.

Instead of immediately working on killing the top-level box, we subdivide in the six co-ordinate directions to get the 64 sub-boxes

$Z(000000), Z(000001), Z(000010), Z(000011), \dots, Z(111111).$

We then throw out the ones with fourth co-ordinate equal to 1 (see condition ‘conjugate’), leaving the 32 sub-boxes

$Z(000000), Z(000001), Z(000010), Z(000011), \dots, Z(111011).$

We then use *verify* to kill these.

The choices in *verify* are made for it by a sequence of integers given as input. The sequence of integers containing the directions for killing $Z(000000)$ is contained in the file `data/000000` (actually, `data/000000.d`). In such a sequence, 0 tells *verify* to subdivide the present box (by $x_i = c$), to position itself on the “left-hand” box ($x_i \leq c$) created by that subdivision, and to read in the next integer in the sequence. A positive integer n tells *verify* to kill directly the sub-box it is positioned at, using the condition (and killerword, if necessary) on line n in the “conditionlist” file, and then to position itself at the “next” natural sub-box. Now, -1 tells *verify* to omit the sub-box, and mark it as skipped (the sequence of integers used in killing the skipped box $Z(s)$ is contained in a file `data/s`).

The checker program *verify*, its inputs, and the list of conditions are available from the *Annals of Mathematics* web site.

FIXME(*Example 5.4.*) To illustrate the checking process in action, this is a (non-representative) example, which shows how the sub-box $Z(s)$ (minus a hole) is killed, where

$$s = 001000110001110111001111000101111111011111 \\ 00111001111000001111011110111.$$

The input associated with this sub-box is

$$(0, 0, 0, 1929, 12304, 0, 0, 7, 0, 1965, 0, 1929, 1929, 1996, -1),$$

which causes the program to kill $Z(s)$ in the following fashion:

```
kill Z(s):
kill Z(s0):
kill Z(s00):
kill Z(s000) with condition 1929 = "L(FwFWFWfWFWFwFwfw)"
kill Z(s001) with condition 12304 = "L(FwfWFFWFwFwfwfWfwfw)"
kill Z(s01):
kill Z(s010):
kill Z(s0100) with condition 7 = "L(w)"
kill Z(s0101):
kill Z(s01010) with condition 1965 = "L(fwFwFWFFWFwFwfwfw)"
kill Z(s01011):
kill Z(s010110) with condition 1929
kill Z(s010111) with condition 1929
kill Z(s011) with condition 1996 = "L(FwFwFWFWfWFWFwFwfw)"
omit Z(s1)
```

as shown in

FIXME(Figure 5.1).

FIGURE 5.1 . Six levels of subdivision, in two projections, with all the trimmings.

$Z(s1)$ is ignored, so the checker would indicate this omission in its report. In fact, $Z(s1)$ is one of the 11 exceptional sub-boxes (seven boxes after joining abutters), specifically X_{5a} , hence killed automatically.

The use of condition “L(w)” so deep in the tree is unusual. In this case, it is because the manifold in the exceptional sub-box has $\text{length}(f) = \text{length}(w)$, so that the program will frequently come to places where it can bound $\text{length}(f) > \text{length}(w)$ nearby.

The sequence 0,1929,1929 in the input for $Z(s)$ tells the checker to subdivide $Z(s01011)$, and then use the condition 1929 on both halves to kill them separately, thereby killing $Z(s01011)$. As mentioned above, the reason for carrying out this subdivision is that the remainder/error bound in the calculation for $Z(s01011)$ using condition 1929 was not good enough to prove that the sub-box is killed directly.

In the input for $Z(s)$ we could have replaced 0,1929,1929 with 1929 alone and then the checker program itself would be smart enough to carry out the subdivision after 1929 failed to kill the sub-box $Z(s01011)$. This recursive subdivision tool is quite useful when dealing with the remainder/error term—if a killerword barely misses killing off a sub-box, then recursively subdivide the sub-box and use the same killerword on the pieces until it succeeds. We note that the theoretical error arising from using a first-order Taylor approximation is likely to be significantly improved by subdivision, whereas the round-off error is relatively unaffected because the killerword used is unchanged (hence the number of mathematical operations performed is unchanged).

The binary numbers used by the computer require too much space to print. In the example calculation which follows, we instead use decimal representations (although we print fewer digits than could be gotten from the 53 binary digits used for the actual calculations).

The sub-box $Z(s01011)$ is the region where

$$\left(\begin{array}{l} -1.381589027741 \dots \leq \operatorname{Re}(L') \leq -1.379848991182 \dots \\ -1.378124546093 \dots \leq \operatorname{Re}(D') \leq -1.376574349753 \dots \\ 0.999893182771 \dots \leq \operatorname{Re}(R') \leq 1.001274250703 \dots \\ -2.535837191243 \dots \leq \operatorname{Im}(L') \leq -2.534606799593 \dots \\ 2.535404997792 \dots \leq \operatorname{Im}(D') \leq -2.534308843448 \dots \\ -0.001953125000 \dots \leq \operatorname{Im}(R') \leq 0.000000000000 \dots \end{array} \right)$$

At this point, we would like to compute

$$f, w, g = f^{-1}w f^{-1}w^{-1} f^{-1}w^{-1} f w^{-1} f^{-1}w^{-1} f^{-1}w f^{-1}w f w w, \operatorname{length}(g),$$

and so on. However, these items take on values over an entire sub-box and thus are computed via `AffApproxes` (first-order Taylor approximations with remainder/error bounds), which are not formally defined until the next section. We complete `FIXME`(Example 5.4) at the end of `FIXME`(Section 6).

`FIXME`(*Remark 5.5*) . For those planning on looking at the program *verify* we now tie in the above description of its workings to a portion of the actual code in the program. We note that the CWeb version of *verify* is extensively documented, and is organized so that the most important details are presented first.

If the executable version of *verify* is called *verify* and we are in the correct place with respect to the location of the data, then a typical UNIX command line would be

```
zcat data/000000.gz | verify 000000 > output000000
```

This would run *verify* at the node 000000, and, when needed, would pipe in the unzipped data from `data/000000.gz`. This unzipped data contains the tree decomposition of the parameter space at the sub-box 000000. The output from *verify* would be redirected to the file `output000000`.

In *verify*, `main` would check for syntax errors in the command line, and if there were no such errors, would read the location 000000 into the character array `where` and compute that the `depth` of `where` was 6, which means that 000000 contains six subdivisions. It would then immediately print

```
verified 000000 - {
```

into the file `output000000`, and then call the function *verify*, as follows:

```
verify(where, depth, 0);
```

The function `verify(where, depth, autocode)` is now invoked; this time with `autocode` equal to 0. *Verify* would first check that `depth` was not too deep. Next, *verify* checks if `autocode` is equal to 0, which it is, so it reads in the next (in this case, the first) integer from the unzipped file `data/000000.gz`,

and sets `code` equal to this integer. Now, *verify* recursively calls itself on the left child (0000000) of the **where** box and the right child (0000001) of the **where** box:

```
where[depth] = '0';
verify(where, depth + 1, code);
where[depth] = '1';
verify(where, depth + 1, code);
```

In general, `verify(where, depth, autocode)` does the following. It checks to see that `depth` is not too deep. Then if `autocode` is equal to 0, it recursively calls itself on its left and right children. If `autocode` is not equal to 0, then `code` is set equal to `autocode`, and three possibilities can occur. Either,

1) `code` is less than zero, in which case we are at a sub-box to be skipped, and *verify* prints out its location (**where**) in `output000000` and recursively moves on to the next node in the tree, or

2) `code` is greater than zero and it invokes a condition/killerword from the file *conditionlist* which kills the entire sub-box **where** in which case *verify* simply recursively moves on to the next node in the tree, or

3) `code` is greater than zero and it invokes a condition/killerword from the file *conditionlist* which does not kill the entire sub-box **where**, in which case *verify* subdivides the sub-box **where** and recursively calls itself on the left child and the right child, using the same `code`:

```
where[depth] = '0';
verify(where, depth + 1, code);
where[depth] = '1';
verify(where, depth + 1, code);
```

In this way *verify* tests the entire starting box, in this case the sub-box 000000, and if successful at killing it minus the omissions which it prints out, it finishes *main* by printing out a right bracket into `output000000`.

4. Affine approximations

Remark 6.1. To show that a sub-box of the parameter box \mathcal{W} is killed by one of the interesting conditions (plus associated killerword) we need to show that at each point in the sub-box, the killerword evaluated at that point satisfies the given condition (see Section 5). That is, we are simply analyzing a certain function from the sub-box to \mathbf{C} .

As described in Remark 6.5, this analysis can be pulled back from the sub-box in question to the closed polydisc $A = \{(z_0, z_1, z_2) \in \mathbf{C}^3 : |z_k| \leq 1 \text{ for } k \in \{0, 1, 2\}\}$. Loosely, we will analyze such a function on A by using Taylor approximations consisting of an affine approximating function together with a bound on the “error” in the approximation (this could also be described as a “remainder bound”). This “error” is separate from round-off error, which will be analyzed in Sections 7 and 8.

Problems 6.2. There are two immediate problems likely to arise from this Taylor approximation approach. The first problem is the appearance of unpleasant functions such as Arccosh . We have already taken care of this problem by “exponentiating” our preliminary parameter space \mathcal{P} . This resulted in all functions under consideration being built up from the co-ordinate functions L', D' , and R' on \mathcal{W} by means of the elementary operations $+$, $-$, \times , $/$, $\sqrt{\cdot}$.

Second, for a given “built-up function” the computer needs to be able to compute the Taylor approximation, and the error term. This will be handled by developing combination formulas for elementary operations (see the propositions below). Specifically, given two Taylor approximations with error terms representing functions g and h and an elementary operation on g and h , we will show how to get the Taylor approximation with error term for the resultant function from the two original Taylor approximations.

A similar approach was developed independently by Stolfi and Figuereido (see [FS]).

Remark 6.3. We set up the Taylor approximation approach rigorously as follows in Definition 6.4. The notation will be a bit unusual, but we are motivated by a desire to stay close to the notation used in the checker computer programs, *verify*. However, it should be pointed out that the formulas in this section will be superseded by the ones in Section 8, which incorporate a round-off error analysis. It is the Section 8 formulas that are used in *verify*.

Definition 6.4. An *AffApprox* x is a five-tuple $(x.f; x.f_0, x.f_1, x.f_2; x.e)$, consisting of four complex numbers $x.f, x.f_0, x.f_1, x.f_2$ and one real number $x.e$, which represents all functions $g : A \rightarrow \mathbf{C}$ such that

$$|g(z_0, z_1, z_2) - (x.f + x.f_0 z_0 + x.f_1 z_1 + x.f_2 z_2)| \leq x.e$$

for all $(z_0, z_1, z_2) \in A$. That is, x represents all functions from A to \mathbf{C} that are $x.e$ -well-approximated by the affine function $x.f + x.f_0 z_0 + x.f_1 z_1 + x.f_2 z_2$. We will denote this set of functions associated with x by $S(x)$.

Remark 6.5. As mentioned in Remark 6.1, given a sub-box to analyze, instead of working with functions defined on the sub-box, we will work with corresponding functions defined on A . Specifically, rather than build up a function by elementary operations performed on the co-ordinate functions L', D', R'

restricted to the given sub-box, we will perform the elementary operations on the following functions defined on A ,

$$(p_0 + ip_3; s_0 + is_3, 0, 0; 0) \quad (p_1 + ip_4; 0, s_1 + is_4, 0; 0) \quad (p_2 + ip_5; 0, 0, s_2 + is_5; 0)$$

where $(p_0 + ip_3, p_1 + ip_4, p_2 + ip_5)$ is the center of the sub-box in question, and the s_i describe the six dimensions of the box. In the computer programs, these three functions are called *along*, *ortho*, and *whirle*, respectively, and p_i and s_i are denoted $pos[i]$ and $size[i]$, respectively.

After the following remarks, we state and prove the combination formulas.

Remarks 6.6. i) In order to co-ordinate numbering with Section 8, we will break with the convention used previously in this paper and start the numbering of the propositions with 6.1. However, we will end this section with Example 6.7.

ii) The negation of a set of functions is the set consisting of the negatives of the original functions, and similarly for other operations.

iii) The propositions that follow include in their statements the definitions of the various operations on AffApproxes. What needs to be proved is that the S functions behave as expected. For example, we need to show that under the definition given for addition, the set of functions $S(x+y)$ contains all functions obtained by adding a function from $S(x)$ to a function from $S(y)$.

PROPOSITION 6.1 (unary minus). *If x is an AffApprox, then $S(-x) = -(S(x))$ where*

$$-x \equiv (-x.f; -x.f_0, -x.f_1, -x.f_2; x.e).$$

Proof.

$$|g(z_0, z_1, z_2) - (x.f + x.f_0 z_0 + x.f_1 z_1 + x.f_2 z_2)| \leq e$$

if and only if

$$|-g(z_0, z_1, z_2) - (-x.f - x.f_0 z_0 - x.f_1 z_1 - x.f_2 z_2)| \leq e. \quad \square$$

PROPOSITION 6.2 (addition). *If x and y are AffApproxes, then $S(x+y) \supseteq S(x) + S(y)$, where*

$$x + y \equiv (x.f + y.f; x.f_0 + y.f_0, x.f_1 + y.f_1, x.f_2 + y.f_2; x.e + y.e).$$

Proof. If $g \in S(x)$ and $h \in S(y)$ then we must show that $g+h \in S(x+y)$.

$$\begin{aligned} & |(g+h)(z_0, z_1, z_2) \\ & - ((x.f + y.f) + (x.f_0 + y.f_0)z_0 + (x.f_1 + y.f_1)z_1 + (x.f_2 + y.f_2)z_2)| \\ & \leq |g(z_0, z_1, z_2) - (x.f + (x.f_0)z_0 + (x.f_1)z_1 + (x.f_2)z_2)| \end{aligned}$$

$$\begin{aligned}
& + |h(z_0, z_1, z_2) - (y.f + (y.f_0)z_0 + (y.f_1)z_1 + (y.f_2)z_2)| \\
& \leq x.e + y.e.
\end{aligned}
\quad \square$$

PROPOSITION 6.3 (subtraction). *If x and y are AffApproxes, then $S(x-y) \supseteq S(x) - S(y)$, where*

$$x - y \equiv (x.f - y.f; x.f_0 - y.f_0, x.f_1 - y.f_1, x.f_2 - y.f_2; x.e + y.e).$$

To co-ordinate numbering with Section 8 (where we incorporate round-off error into these formulas) we include the following special cases of Propositions 6.2 and 6.3. Similarly for Propositions 6.7, 6.9, and 6.10.

In what follows, “double” refers to a real number, and has an associated AffApprox, with the last four entries zero. When we do machine arithmetic in Sections 7 and 8, doubles will be machine numbers.

PROPOSITION 6.4 (addition of an AffApprox and a double). *If x is an AffApprox and y is a double, then $S(x + y) \supseteq S(x) + S(y)$, where*

$$x + y \equiv (x.f + y; x.f_0, x.f_1, x.f_2; x.e).$$

PROPOSITION 6.5 (subtraction of a double from an AffApprox). *If x is an AffApprox and y is a double, then $S(x - y) \supseteq S(x) - S(y)$, where*

$$x - y \equiv (x.f - y; x.f_0, x.f_1, x.f_2; x.e).$$

PROPOSITION 6.6 (multiplication). *If x and y are AffApproxes, then $S(x \times y) \supseteq S(x) \times S(y)$, where*

$$\begin{aligned}
x \times y \equiv & (x.f \times y.f; x.f \times y.f_0 + x.f_0 \times y.f, \\
& x.f \times y.f_1 + x.f_1 \times y.f, x.f \times y.f_2 + x.f_2 \times y.f; \\
& (\text{size}(x) + x.e) \times (\text{size}(y) + y.e) + (|x.f| \times y.e + x.e \times |y.f|))
\end{aligned}$$

with $\text{size}(x) = |x.f_0| + |x.f_1| + |x.f_2|$ and $\text{size}(y) = |y.f_0| + |y.f_1| + |y.f_2|$.

Proof. If $g \in S(x)$ and $h \in S(y)$ then we must show that $g \times h \in S(x \times y)$. That is, we need to show

$$\begin{aligned}
& |(g \times h)(z_0, z_1, z_2) - ((x.f \times y.f) + (x.f \times y.f_0 + x.f_0 \times y.f)z_0 \\
& + (x.f \times y.f_1 + x.f_1 \times y.f)z_1 + (x.f \times y.f_2 + x.f_2 \times y.f)z_2)| \\
& \leq (\text{size}(x) + x.e) \times (\text{size}(y) + y.e) + (|x.f| \times y.e + x.e \times |y.f|).
\end{aligned}$$

Note that for any point $(z_0, z_1, z_2) \in A$ and any functions $g \in S(x)$ and $h \in S(y)$ we can find complex numbers u, v with $|u| \leq 1$ and $|v| \leq 1$, such that

$$g(z_0, z_1, z_2) = x.f + (x.f_0 z_0 + x.f_1 z_1 + x.f_2 z_2) + (x.e)u$$

and

$$h(z_0, z_1, z_2) = y.f + (y.f_0 z_0 + y.f_1 z_1 + y.f_2 z_2) + (y.e)v.$$

Multiplying out, we see that

$$\begin{aligned} (g \times h)(z_0, z_1, z_2) &= (x.f \times y.f) + (x.f \times y.f_0 + x.f_0 \times y.f)z_0 \\ &\quad + (x.f \times y.f_1 + x.f_1 \times y.f)z_1 + (x.f \times y.f_2 + x.f_2 \times y.f)z_2 \\ &\quad + (x.f \times y.e)v + (x.e \times y.f)u + ((x.f_0 z_0 + x.f_1 z_1 + x.f_2 z_2) + (x.e)u) \\ &\quad \times ((y.f_0 z_0 + y.f_1 z_1 + y.f_2 z_2) + (y.e)v). \end{aligned}$$

Hence,

$$\begin{aligned} &|(g \times h)(z_0, z_1, z_2) - ((x.f \times y.f) \\ &\quad + ((x.f \times y.f_0 + x.f_0 \times y.f)z_0 \\ &\quad + (x.f \times y.f_1 + x.f_1 \times y.f)z_1 + (x.f \times y.f_2 + x.f_2 \times y.f)z_2))| \\ &\leq (|x.f|y.e + x.e|y.f|) + (\text{size}(x) + x.e) \times (\text{size}(y) + y.e). \end{aligned} \quad \square$$

PROPOSITION 6.7 (an AffApprox multiplied by a double). *If x is an AffApprox and y is a double, then $S(x \times y) \supseteq S(x) \times S(y)$, where*

$$x \times y \equiv (x.f \times y; x.f_0 \times y, x.f_1 \times y, x.f_2 \times y; x.e \times |y|).$$

PROPOSITION 6.8 (division). *If x and y are AffApproxes with $|y.f| > \text{size}(y) + y.e$, then $S(x/y) \supseteq S(x)/S(y)$, where*

$$\begin{aligned} x/y &\equiv (x.f/y.f; (-x.f \times y.f_0 + x.f_0 \times y.f)/((y.f)^2), \\ &\quad (-x.f \times y.f_1 + x.f_1 \times y.f)/((y.f)^2), \\ &\quad (-x.f \times y.f_2 + x.f_2 \times y.f)/((y.f)^2); \\ &\quad (|x.f| + \text{size}(x) + x.e)/(|y.f| - (\text{size}(y) + y.e)) \\ &\quad - ((|x.f|/|y.f| + \text{size}(x)/|y.f|) + |x.f|\text{size}(y)/(|y.f||y.f|))). \end{aligned}$$

Proof. For notational convenience, denote $(x.f_0 z_0 + x.f_1 z_1 + x.f_2 z_2)$ by $x.f_k z_k$ and similarly for $y.f_k z_k$ and so on. As above, note that for any point $(z_0, z_1, z_2) \in A$ and any functions $g \in S(x)$ and $h \in S(y)$ we can find complex numbers u, v with $|u| \leq 1$ and $|v| \leq 1$, such that

$$g(z_0, z_1, z_2) = x.f + (x.f_k z_k) + (x.e)u$$

and

$$h(z_0, z_1, z_2) = y.f + (y.f_k z_k) + (y.e)v.$$

We compare $(g/h)(z_0, z_1, z_2)$ with its putative affine approximation. That is, we analyze

$$\left| (x.f + (x.f_k z_k) + (x.e)u) / (y.f + (y.f_k z_k) + (y.e)v) - ((x.f/y.f) + \frac{(x.f_k)y.f - x.f(y.f_k)}{(y.f)^2} z_k) \right|.$$

Putting this over a common denominator of $|((y.f)^2)(y.f + (y.f_k z_k) + (y.e)v)|$ and cancelling equal terms (in the numerator) we are left with a quotient whose numerator is

$$\begin{aligned} & |x.e((y.f)^2)u - (x.f_k)y.f(y.f_k)z_k - x.f((y.f_k)^2)z_k \\ & + (x.f)y.f(y.e)v + x.f_k(y.f)y.e(v)z_k - x.f(y.f_k)y.e(v)z_k|. \end{aligned}$$

We must show this (first) quotient is bounded by

$$\begin{aligned} & (|x.f| + \text{size}(x) + x.e) / (|y.f| - (\text{size}(y) + y.e)) \\ & - ((|x.f|/|y.f| + \text{size}(x)/|y.f|) + |x.f|\text{size}(y)/(|y.f||y.f|)). \end{aligned}$$

Putting this over a common denominator of $|y.f|^2(|y.f| - (\text{size}(y) + y.e))$ and cancelling equal terms (in the numerator) we are left with a second quotient, whose numerator is

$$x.e|y.f|^2 - (-|x.f||y.f|y.e - \text{size}(x)|y.f|(\text{size}(y) + y.e) - |x.f|\text{size}(y)(\text{size}(y) + y.e))$$

and we see that all terms in this numerator are positive. Further, the terms in the numerators of the first and second quotients correspond in a natural way, and each term in the numerator of the second quotient is greater than or equal to the absolute value of its corresponding term in the numerator of the first quotient.

Finally, because the denominator in the second quotient is less than or equal to the absolute value of the denominator in the first quotient, we see that the absolute value of the first quotient is less than or equal to the second quotient, as desired. \square

PROPOSITION 6.9 (division of a double by an AffApprox). *If x is a double and y is an AffApprox with $|y.f| > \text{size}(y) + y.e$, then $S(x/y) \supseteq S(x)/S(y)$, where*

$$\begin{aligned} x/y \equiv & (x/y.f; -x \times y.f_0/((y.f)^2), -x.f \times y.f_1/((y.f)^2), -x.f \times y.f_2/((y.f)^2); \\ & (|x|/(|y.f| - (\text{size}(y) + y.e)) - (|x|/|y.f| + |x|\text{size}(y)/(|y.f||y.f|))). \end{aligned}$$

PROPOSITION 6.10 (division of an AffApprox by a double). *If x is an AffApprox and y is a double with $|y| > 0$, then $S(x/y) \supseteq S(x)/S(y)$, where*

$$x/y \equiv (x.f/y; x.f_0/y, x.f_1/y, x.f_2/y; x.e/|y|).$$

Finally, we do the square root.

PROPOSITION 6.11 (square root). *If x is an AffApprox with $|x.f| > \text{size}(x) + x.e$, then $S(\sqrt{x}) \supseteq \sqrt{S(x)}$, where*

$$\begin{aligned} \sqrt{x} = & \left(\sqrt{x.f}; \frac{x.f_0}{2\sqrt{x.f}}, \frac{x.f_1}{2\sqrt{x.f}}, \frac{x.f_2}{2\sqrt{x.f}}; \right. \\ & \left. \sqrt{|x.f|} - \left(\frac{\text{size}(x)}{2\sqrt{|x.f|}} + \sqrt{|x.f| - (\text{size}(x) + x.e)} \right) \right). \end{aligned}$$

If $|x.f| \leq \text{size}(x) + x.e$ then we use the crude estimate

$$(0; 0, 0, 0; \sqrt{|x.f| + \text{size}(x) + x.e}).$$

The branch of the square root of a complex number is determined by the construction of the square root of a complex in Proposition 7.14. In fact, the square root is in the first or fourth quadrant.

Proof. As above, note that for any point $(z_0, z_1, z_2) \in A$ and any function $g \in S(x)$ we can find a complex number u with $|u| \leq 1$, such that

$$g(z_0, z_1, z_2) = x.f + (x.f_k z_k) + (x.e)u.$$

Also, because $|x.f| > \text{size}(x) + x.e$, we see that the argument of $x.f + (x.f_k z_k) + (x.e)u$ is within $\pi/2$ of the argument of $x.f$, and therefore, we can require that $\sqrt{g(z_0, z_1, z_2)}$ have argument within $\pi/4$ of the argument of $\sqrt{x.f}$.

We need to show that

$$\begin{aligned} & \left| \sqrt{x.f + x.f_k z_k + (x.e)u} - \left(\sqrt{x.f} + \frac{x.f_k z_k}{2\sqrt{x.f}} \right) \right| \\ & \leq \sqrt{|x.f|} - \left(\frac{\text{size}(x)}{2\sqrt{|x.f|}} + \sqrt{|x.f| - (\text{size}(x) + x.e)} \right). \end{aligned}$$

Or, after we multiply both sides by $\sqrt{|x.f|}$,

$$\begin{aligned} & \left| \sqrt{x.f(x.f + x.f_k z_k + (x.e)u)} - (x.f + (x.f_k)z_k/2) \right| \\ & \leq (|x.f| - \text{size}(x)/2) - \sqrt{|x.f|(|x.f| - (\text{size}(x) + x.e))}. \end{aligned}$$

The two sides of the inequality are of the form $A - B$ and $C - D$, and we “simplify” by multiplying by $\frac{A+B}{A+B}$ and $\frac{C+D}{C+D}$. We now show that the (absolute value of the) left-hand numerator is less than or equal to the right-hand numerator. Later, we will show that the (absolute value of the) left-hand denominator is larger than or equal to the right-hand denominator. The left-hand numerator

is

$$\begin{aligned}
& |x.f(x.f + x.f_k z_k + (x.e)u) - (x.f + (x.f_k)z_k/2)^2| \\
&= |(x.f)^2 + x.f(x.f_k)z_k + x.f(x.e)u - (x.f)^2 \\
&\quad - x.f(x.f_k)z_k - ((x.f_k)^2)(z_k)^2/4| \\
&= |x.f(x.e)u - ((x.f_k)^2)(z_k)^2/4|.
\end{aligned}$$

The right-hand numerator is

$$\begin{aligned}
& (|x.f| - \text{size}(x)/2)^2 - |x.f|(|x.f| - (\text{size}(x) + x.e)) \\
&= |x.f|^2 - |x.f|\text{size}(x) + \text{size}(x)^2/4 - |x.f|^2 + |x.f|\text{size}(x) + |x.f|x.e \\
&= |x.f|x.e + \text{size}(x)^2/4.
\end{aligned}$$

So the left-hand numerator is indeed less than or equal to the right-hand numerator.

We now compare the denominators, but only after dividing each by $\sqrt{|x.f|}$. The left-hand denominator is

$$\left| \sqrt{x.f + x.f_k z_k + (x.e)u} + \left(\sqrt{x.f} + \frac{x.f_k z_k}{2\sqrt{x.f}} \right) \right|$$

while the right-hand denominator is

$$\sqrt{|x.f|} - \frac{\text{size}(x)}{2\sqrt{|x.f|}} + \sqrt{|x.f| - (\text{size}(x) + x.e)}.$$

The claim that the left-hand denominator is greater than or equal to the right-hand denominator is a bit complicated. First, compare the $\sqrt{x.f}$ term and the $\sqrt{|x.f|}$ terms. They are the same distance from the origin. Next, note that as z_k and u take on all relevant values, $x.f + x.f_k z_k + (x.e)u$ describes a disk centered at $x.f$ with radius less than $|\sqrt{x.f}|$. Hence, $\sqrt{x.f + x.f_k z_k + (x.e)u}$ describes a convex set containing $\sqrt{x.f}$. This set is symmetric about the line joining the origin and $\sqrt{x.f}$. Further, $\sqrt{x.f} + \sqrt{x.f + x.f_k z_k + (x.e)u}$ describes a convex set containing $2\sqrt{x.f}$. This set is also symmetric about the line joining the origin and $\sqrt{x.f}$. It is easy enough to see that no points on this convex symmetric set get closer to the origin than $\sqrt{|x.f|} + \sqrt{|x.f|} - (\text{size}(x) + x.e)$.

Finally, because $|\frac{x.f_k z_k}{2\sqrt{x.f}}| \leq \frac{\text{size}(x)}{2\sqrt{|x.f|}}$, no points of

$$\sqrt{x.f} + \sqrt{x.f + x.f_k z_k + (x.e)u} + \frac{x.f_k z_k}{2\sqrt{x.f}}$$

can get closer to the origin than

$$\sqrt{|x.f|} + \sqrt{|x.f|} - (\text{size}(x) + x.e) - \frac{\text{size}(x)}{2\sqrt{|x.f|}}. \quad \square$$

Example 6.7 (Continuation of Example 5.4). We can now complete the analysis begun in Example 5.4, because we can describe f and w as 2-by-2 matrices of AffApproxes. We note the minor quibble that the full definition of AffApprox is given in Section 8, where round-off error is incorporated into the remainder/error-bound term.

For convenience, we repeat the description of the sub-box under investigation. The sub-box $Z(s01011)$ with

$$s = 00100011000111011100111100010111111101111100111001111000001111011110111$$

is the region where

$$\left(\begin{array}{l} -1.381589027741 \dots \leq \text{Re}(L') \leq -1.379848991182 \dots \\ -1.378124546093 \dots \leq \text{Re}(D') \leq -1.376574349753 \dots \\ 0.999893182771 \dots \leq \text{Re}(R') \leq 1.001274250703 \dots \\ -2.535837191243 \dots \leq \text{Im}(L') \leq -2.534606799593 \dots \\ 2.535404997792 \dots \leq \text{Im}(D') \leq -2.534308843448 \dots \\ -0.001953125000 \dots \leq \text{Im}(R') \leq 0.000000000000 \dots \end{array} \right).$$

For this sub-box, we get (printing only 10 decimal places, for visual convenience):

$$f = \left[\begin{array}{c} \left(\begin{array}{c} -0.8677851121 + i1.4607429651; \\ 0.0000248810 - i0.0003125810, \\ 0.0000000000 + i0.0000000000, \\ 0.0000000000 + i0.0000000000; \\ 0.0000000289 \\ 0.0000000000 + i0.0000000000; \\ 0.0000000000 + i0.0000000000, \\ 0.0000000000 + i0.0000000000, \\ 0.0000000000 + i0.0000000000; \\ 0.0000000000 \end{array} \right) \\ \left(\begin{array}{c} 0.0000000000 + i0.0000000000; \\ 0.0000000000 + i0.0000000000, \\ 0.0000000000 + i0.0000000000, \\ 0.0000000000 + i0.0000000000; \\ 0.0000000000 \end{array} \right) \end{array} \right] \left[\begin{array}{c} \left(\begin{array}{c} 0.0000000000 + i0.0000000000; \\ 0.0000000000 + i0.0000000000, \\ 0.0000000000 + i0.0000000000, \\ 0.0000000000 + i0.0000000000; \\ 0.0000000000 \\ -0.3006023265 - i0.5060039953; \\ -0.0000909686 - i0.0000593570, \\ 0.0000000000 + i0.0000000000, \\ 0.0000000000 + i0.0000000000; \\ 0.0000000301 \end{array} \right) \end{array} \right]$$

and

$$w = \left[\begin{array}{c} \left(\begin{array}{c} -0.5845111829 + i0.4773282853; \\ 0.0000000000 + i0.0000000000, \\ -0.0000296707 - i0.0001657332, \\ -0.0004345111 - i0.0001209539; \\ 0.0000000259 \\ -0.2832291572 + i0.9833572297; \\ 0.0000000000 + i0.0000000000, \\ 0.0000515806 - i0.0001129408, \\ -0.0005778031 + i0.0002005440; \\ 0.0000000286 \end{array} \right) \\ \left(\begin{array}{c} -0.2840228472 + i0.9825063583; \\ 0.0000000000 + i0.0000000000, \\ 0.0000516606 - i0.0001128245, \\ 0.0005776611 - i0.0001998632; \\ 0.0000000642 \\ -0.5846352333 + i0.4764792236; \\ 0.0000000000 + i0.0000000000, \\ -0.0000294917 - i0.0001656653, \\ 0.0004341392 + i0.0001213070; \\ 0.00000005286 \end{array} \right) \end{array} \right].$$

Calculation of $g = f^{-1}wf^{-1}w^{-1}f^{-1}w^{-1}fw^{-1}f^{-1}w^{-1}f^{-1}wf^{-1}wfw$ gives

$$g = \left[\begin{array}{c} \left(\begin{array}{c} -0.5764337542 + i0.4752708071; \\ -0.0031657223 - i0.0001436786, \\ -0.0017723577 + i0.0000352928, \\ -0.0011623491 + i0.0017516088; \\ 0.0008229225 \\ -0.2861207992 + i0.9766064999; \\ -0.0002777968 + i0.0020330488, \\ 0.0000837571 + i0.0010241875, \\ 0.0028322367 - i0.0005972336; \\ 0.0018172437 \end{array} \right) \\ \left(\begin{array}{c} -0.2704033973 + i0.9822741250; \\ -0.0045902952 - i0.0019135041, \\ -0.0026219461 - i0.0007506230, \\ -0.0002823450 + i0.0033805602; \\ 0.0008037640 \\ -0.5861133046 + i0.4624368851; \\ -0.0021932627 + i0.0040523411, \\ -0.0008612361 + i0.0022394639, \\ 0.0061581377 - i0.0005862070; \\ 0.0017738513 \end{array} \right) \end{array} \right].$$

We then get

$$\text{length}(g) = \begin{pmatrix} -1.3588762105 - i2.4897230182; \\ 0.0030210500 - i0.0182284729, \\ 0.0007938572 - i0.0096614614, \\ -0.0122034521 + i0.0074353043; \\ 0.0080071969 \end{pmatrix}$$

and

$$\frac{\text{length}(g)}{L'} = \begin{pmatrix} 0.9825397896 - i0.0008933519; \\ 0.0053701602 + i0.0037789019, \\ 0.0028076072 + i0.0018421952, \\ -0.0002400615 - i0.0049443045; \\ 0.0027802966 \end{pmatrix}.$$

This is not quite good enough to kill the sub-box, since $|\text{length}(g)/L'|$ can be as high as 1.0001951323.

When we subdivide $Z(s01011)$, we have to analyze two sub-boxes, $Z(s010110)$ and $Z(s010111)$. For $Z(s010110)$, the same calculation on the region

$$\begin{aligned} -1.381589027741073400 &\leq \text{Re}(L') \leq -1.379848991182205200 \\ -1.378124546093485700 &\leq \text{Re}(D') \leq -1.376574349753672900 \\ 0.999893182771602220 &\leq \text{Re}(R') \leq 1.001274250703607400 \\ -2.535837191243490300 &\leq \text{Im}(L') \leq -2.534606799593201600 \\ -2.535404997792558600 &\leq \text{Im}(D') \leq -2.534308843448505900 \\ -0.001953125000000000 &\leq \text{Im}(R') \leq -0.000976562500000000 \end{aligned}$$

gives

$$\frac{\text{length}(g)}{L'} = \begin{pmatrix} 0.9814518667 + i0.0008103446; \\ 0.0053616729 + i0.0037834001, \\ 0.0028027236 + i0.0018435245, \\ -0.0013175066 - i0.0032448794; \\ 0.0019033926 \end{pmatrix},$$

and we can then bound $\left| \frac{\text{length}(g)}{L'} \right| \leq 0.9967745579$, which kills $Z(s010110)$.

On $Z(s010111)$, the calculation gives

$$\frac{\text{length}(g)}{L'} = \begin{pmatrix} 0.9836225919 - i0.0025990177; \\ 0.0053786346 + i0.0037743930, \\ 0.0028124892 + i0.0018408583, \\ -0.0013333182 - i0.0032343347; \\ 0.0019044429 \end{pmatrix}$$

and $\left| \frac{\text{length}(g)}{L'} \right| \leq 0.9989610507$, which kills $Z(s010111)$.

5. Complex numbers with round-off error

Remark 7.1. The theoretical method for proving Theorem 0.2 has been implemented via the computer program *verify*, which is available, together with the relevant data sets, at the *Annals* web site. To make this computer-aided proof rigorous, we needed to deal with round-off error in calculations.

One approach to round-off error would be to use interval arithmetic packages to carry out all calculations with floating-point machine numbers, or to generate our own version of these packages. However, it appears that this approach would be much too slow given the size of our collection of sub-boxes and conditions/killerwords.

To solve this problem of speed, we implement round-off error at a higher level of programing. That is, we incorporate round-off error directly into *AffApproxes*, which makes our error calculations more accurate, thereby avoiding much subdivision of sub-boxes. This necessitates that we incorporate round-off error directly into complex numbers as well. In this section we show how to do standard operations on complex numbers while keeping track of round-off error. In the next section we work with *AffApproxes*.

Definition 7.2. There are two types of complex numbers to consider:

- 1) An *XComplex* $x = (x.\text{re}, x.\text{im})$ corresponds to a complex number that is represented exactly; it simply consists of a real part and an imaginary part.
- 2) An *ACComplex* $x = (x.\text{re}, x.\text{im}; x.e)$ corresponds to an “interval” that contains the complex number in question. Thus, it consists of an *XComplex* and a floating-point number representing the error. In particular, the *ACComplex* x represents the set $S(x)$ of complex numbers $\{w : |w - (x.\text{re} + i(x.\text{im}))| \leq x.e\}$. Note that $S(x)$ is also defined for an *XComplex* if we conceptualize an *XComplex* as an *ACComplex* with $x.e = 0$.

Remark 7.3. In general, our operations act on XComplexes and produce AComplexes, or they act on AComplexes and produce AComplexes. In one case, the unary minus, an XComplex goes to an XComplex. In the calculations that follow the effect on the error is the whole point.

Conventions and standards 7.4. We begin, by writing down our basic rules, which follow easily from the IEEE-754 standard for machine arithmetic (see [IEEE]). (Actually, the “hypot” function $h(a, b)$, which computes by elaborate chicanery $\sqrt{a^2 + b^2}$, is not part of the IEEE-754 standard, but satisfies the appropriate standard according to the documentation provided (see [K1]).) The operations here are on double-precision floating-point real numbers (“doubles”) and we denote a true operation by the usual symbol and the associated machine operation by the same symbol in a circle, with two exceptions: a machine square root \sqrt{a} is denoted $\sqrt[4]{a}$ and the machine version of the hypot function is denoted h_{\circ} . Perhaps a third exception is our occasional notation of true multiplication by the absence of a symbol.

There is a finite set of numbers (sometimes called “machine numbers”) which are representable on the computer. With technicalities ignored, a nonzero floating-point number is represented by a fixed number of bits of which the first determines the sign of the number, the next m represent the exponent, and the remaining n represent the mantissa of the number. Because our nonzero numbers start with a 1, that means the n mantissa bits actually represent the next n binary digits after the 1. That is, the mantissa is actually $1.b_1b_2b_3\dots b_n$. The IEEE-754 standard calls for 64-bit doubles with $m = 11$ and $n = 52$. We define EPS to be 2^{-n} , in which case $EPS/2$ is $2^{-(n+1)}$.

The IEEE-754 standard states that the result of an operation is always the closest representable number to the true solution (as long as we are in the bounds of representable numbers). For example, for machine numbers a and b , we have $a \oplus b = m(a + b)$ where m is the function which takes the machine value of its argument (when it lies in the range of representable numbers). Thus, properties of the type

$$|(a + b) - (a \oplus b)| \leq (EPS/2)|a + b|$$

follow immediately from the IEEE-754 standard, as long as we do not *underflow* or *overflow* outside of the range of representable numbers. Specifically, underflow occurs when the result of an operation is smaller in absolute value than 2^{-1022} , and overflow occurs when the result of an operation is larger in absolute value than roughly 2^{1024} (see [IEEE, §7]).

We further note that the formula

$$|(a + b) - (a \oplus b)| \leq (EPS/2)|a \oplus b|$$

follows because the true answer has “exponent” which is less than or equal to the exponent of the machine answer. We reiterate, that in both cases, a and b are assumed to be machine numbers.

Of course, a machine operation such as \oplus must act on doubles, while a “true” operation such as $+$ can act on reals (which includes doubles). In this chapter, long strings of inequalities will be used to prove the various propositions, and care was taken to ensure that machine operations act on machine numbers. In particular, the various variables appearing in the propositions are assumed to be doubles. The IEEE-754 standard provides for conversions from decimal to binary (within the appropriate range, conversion is to the nearest representable number) and from binary to decimal. However, these are rarely used in this paper, although a trivial class of exceptions is provided by the decimal numbers in the conditions of Section 5.

When calculations underflow or overflow outside of the range of representable numbers, we require that the computer inform us if either exception has occurred.

As in Section 6, we now break with the usual numbering convention. Note that the above comments provide a proof of the following properties.

Basic Properties 7.0 (assuming no underflow and no overflow).

In the formulas that follow, a, b , and A are machine numbers and $1 + k \times EPS = 1 \oplus (k \otimes EPS)$ when k is an integer which is not huge in absolute value (that is, smaller than roughly 2^{50}). Thus, within the appropriate range, $1 + k \times EPS$ is a machine number. Similarly, $2^k \times A = 2^k \otimes A$ when k is an integer and $2^k \otimes A$ neither underflows nor overflows.

$$\begin{aligned}
|(a + b) - (a \oplus b)| &\leq (EPS/2)|a + b|, \\
|(a + b) - (a \oplus b)| &\leq (EPS/2)|a \oplus b|, \\
|(a - b) - (a \ominus b)| &\leq (EPS/2)|a - b|, \\
|(a - b) - (a \ominus b)| &\leq (EPS/2)|a \ominus b|, \\
|(a \times b) - (a \otimes b)| &\leq (EPS/2)|a \times b|, \\
|(a \times b) - (a \otimes b)| &\leq (EPS/2)|a \otimes b|, \\
|(a/b) - (a \oslash b)| &\leq (EPS/2)|a/b|, \\
|(a/b) - (a \oslash b)| &\leq (EPS/2)|a \oslash b|, \\
|\sqrt{a} - \sqrt[3]{a}| &\leq (EPS/2)|\sqrt{a}|, \\
|\sqrt{a} - \sqrt[3]{a}| &\leq (EPS/2)|\sqrt[3]{a}|, \\
|h(a, b) - h_{\circ}(a, b)| &\leq (EPS)|h(a, b)|, \\
|h(a, b) - h_{\circ}(a, b)| &\leq (EPS)|h_{\circ}(a, b)|.
\end{aligned}$$

From these formulas, we immediately compute the following.

$$\begin{aligned} (1 - EPS/2)|a + b| &\leq |a \oplus b| \leq (1 + EPS/2)|a + b|, \\ (1 - EPS/2)|a \oplus b| &\leq |a + b| \leq (1 + EPS/2)|a \oplus b|, \end{aligned}$$

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$$\begin{aligned} (1 - EPS/2)|\sqrt{a}| &\leq |\sqrt[3]{a}| \leq (1 + EPS/2)|\sqrt{a}|, \\ (1 - EPS/2)|\sqrt[3]{a}| &\leq |\sqrt{a}| \leq (1 + EPS/2)|\sqrt[3]{a}|, \\ (1 - EPS)|h(a, b)| &\leq |h_{\circ}(a, b)| \leq (1 + EPS)|h(a, b)|, \\ (1 - EPS)|h_{\circ}(a, b)| &\leq |h(a, b)| \leq (1 + EPS)|h_{\circ}(a, b)|. \end{aligned}$$

Of course, we can also get the following type of formula, which is sometimes convenient, for example, in the proof of Lemma 7.2:

$$\left(\frac{1}{1 + \frac{EPS}{2}} \right) |a \oplus b| \leq |a + b| \leq \left(\frac{1}{1 - \frac{EPS}{2}} \right) |a \oplus b|.$$

Before stating our propositions, we prove two lemmas.

LEMMA 7.0 (assuming no underflow and no overflow). *For machine numbers a and b ,*

$$(1 - EPS) \otimes |a \oplus b| \leq |a + b| \leq (1 + EPS) \otimes |a \oplus b|.$$

Analogous formulas hold for $-$, $$, $/$, $\sqrt{\cdot}$.*

Proof. Assume $a + b > 0$. If $(1 + EPS) \otimes (a \oplus b) < (a + b)$ then the machine number $(1 + EPS) \otimes (a \oplus b)$ is a better approximation to $a + b$ than $a \oplus b$, because $(a \oplus b) < (1 + EPS) \otimes (a \oplus b)$. This contradicts the IEEE standard. The case $a + b < 0$ can be handled similarly, and the case $a + b = 0$ is trivial, similarly for the left-hand inequality. \square

LEMMA 7.1.

$$(1 + EPS/2)^a A \leq (1 + kEPS) \otimes A$$

where A is a nonnegative machine number, and a is a (not huge) integer, such that for a even, $k = \frac{a}{2} + 1$ and for a odd, $k = \frac{a+1}{2} + 1$.

Proof.

$$(1 + EPS/2)^a A \leq (1 - EPS/2)(1 + kEPS)A \leq (1 + kEPS) \otimes A.$$

The first inequality holds if a and k are as in the lemma, and the second inequality is a consequence of one of the formulas preceding Lemma 7.0 ($A \geq 0$). \square

We now begin our construction of complex arithmetic. We will give proofs for most of the operations; the others should be straightforward to derive, or can be found in the *Annals* web site.

Remarks 7.5. i) We remind the reader that all machine operations are on machine numbers, and that the various variables appearing in the propositions are assumed to be doubles.

ii) The propositions that follow include in their statements the definitions of the various operations (see Remark 6.6iii).

PROPOSITION 7.1 ($-X$). *If x is an XComplex, then*

$$-x \equiv (-x.\text{re}, -x.\text{im}).$$

PROPOSITION 7.2 ($X+D$). *If x is an XComplex and d is a double, then $S(x+d) \supseteq S(x) + S(d)$, where*

$$x + d \equiv (x.\text{re} \oplus d, x.\text{im}; (EPS/2) \otimes |x.\text{re} \oplus d|).$$

Proof. The error is bounded by

$$|(x.\text{re} + d) - (x.\text{re} \oplus d)| \leq (EPS/2)|x.\text{re} \oplus d| = (EPS/2) \otimes |x.\text{re} \oplus d|. \quad \square$$

PROPOSITION 7.3 ($X-D$). *If x is an XComplex and d is a double, then $S(x-d) \supseteq S(x) - S(d)$, where*

$$x - d \equiv (x.\text{re} \ominus d, x.\text{im}; (EPS/2) \otimes |x.\text{re} \ominus d|).$$

PROPOSITION 7.4 ($X+X$). *If x and y are XComplexes, then $S(x+y) \supseteq S(x) + S(y)$, where*

$$\begin{aligned} x + y &\equiv (x.\text{re} \oplus y.\text{re}, x.\text{im} \oplus y.\text{im}; (EPS/2) \\ &\quad \otimes ((1 + EPS) \otimes (|x.\text{re} \oplus y.\text{re}| \oplus |x.\text{im} \oplus y.\text{im}|))). \end{aligned}$$

Proof. The error is bounded by

$$\begin{aligned} &|(x.\text{re} + y.\text{re}) - (x.\text{re} \oplus y.\text{re})| + |(x.\text{im} + y.\text{im}) - (x.\text{im} \oplus y.\text{im})| \\ &\leq (EPS/2)(|x.\text{re} \oplus y.\text{re}| + |x.\text{im} \oplus y.\text{im}|) \\ &\leq (EPS/2)((1 + EPS) \otimes (|x.\text{re} \oplus y.\text{re}| \oplus |x.\text{im} \oplus y.\text{im}|)) \\ &= (EPS/2) \otimes ((1 + EPS) \otimes (|x.\text{re} \oplus y.\text{re}| \oplus |x.\text{im} \oplus y.\text{im}|)). \end{aligned}$$

To go from line 2 to line 3 we used Lemma 7.0. \square

PROPOSITION 7.5 ($X - X$). *If x and y are XComplexes, then $S(x - y) \supseteq S(x) - S(y)$, where*

$$\begin{aligned} x - y &\equiv (x.\text{re} \ominus y.\text{re}, x.\text{im} \ominus y.\text{im}; \\ &\quad (EPS/2) \otimes ((1 + EPS) \otimes (|x.\text{re} \ominus y.\text{re}| \oplus |x.\text{im} \ominus y.\text{im}|))). \end{aligned}$$

PROPOSITION 7.6 ($A + A$). *If x and y are AComplexes, then $S(x + y) \supseteq S(x) + S(y)$, where*

$$\begin{aligned} x + y &\equiv (\text{re}, \text{im}; e) \text{ with} \\ \text{re} &= x.\text{re} \oplus y.\text{re} \\ \text{im} &= x.\text{im} \oplus y.\text{im} \\ e &= (1 + 2EPS) \otimes (((EPS/2) \otimes (|\text{re}| \oplus |\text{im}|)) \oplus (x.e \oplus y.e)). \end{aligned}$$

Proof. The error is bounded by the sum of the contributions from the real part, the imaginary part, and the two individual errors:

$$\begin{aligned} &|(x.\text{re} \oplus y.\text{re}) - (x.\text{re} + y.\text{re})| + |(x.\text{im} \oplus y.\text{im}) - (x.\text{im} + y.\text{im})| + (x.e + y.e). \\ &\leq (EPS/2)|x.\text{re} \oplus y.\text{re}| + (EPS/2)|x.\text{im} \oplus y.\text{im}| + (1 + EPS/2)(x.e \oplus y.e) \\ &\leq (1 + EPS/2)(EPS/2)(|x.\text{re} \oplus y.\text{re}| \oplus |x.\text{im} \oplus y.\text{im}|) \\ &\quad + (1 + EPS/2)(x.e \oplus y.e) \\ &= (1 + EPS/2)((EPS/2)(|x.\text{re} \oplus y.\text{re}| \oplus |x.\text{im} \oplus y.\text{im}|) + (x.e \oplus y.e)) \\ &\leq (1 + EPS/2)^2(((EPS/2)(|x.\text{re} \oplus y.\text{re}| \oplus |x.\text{im} \oplus y.\text{im}|)) \oplus (x.e \oplus y.e)) \\ &\leq (1 + 2EPS) \otimes (((EPS/2) \otimes (|x.\text{re} \oplus y.\text{re}| \oplus |x.\text{im} \oplus y.\text{im}|)) \oplus (x.e \oplus y.e)). \end{aligned}$$

□

The precedence for machine operations is the same as that for true operations, so some parentheses are unnecessary and will often be omitted in what follows.

PROPOSITION 7.7 ($A - A$). *If x and y are AComplexes, then $S(x - y) \supseteq S(x) - S(y)$, where*

$$\begin{aligned} x - y &\equiv (\text{re}, \text{im}; e) \text{ with} \\ \text{re} &= x.\text{re} \ominus y.\text{re} \\ \text{im} &= x.\text{im} \ominus y.\text{im} \\ e &= (1 + 2EPS) \otimes (((EPS/2) \otimes (|\text{re}| \oplus |\text{im}|)) \oplus (x.e \oplus y.e)). \end{aligned}$$

PROPOSITION 7.8 ($X \times D$). *If x is an XComplex and d is a double, then $S(x \times d) \supseteq S(x) \times S(d)$, where*

$$\begin{aligned} x \times d &\equiv (\text{re}, \text{im}; e) \text{ with} \\ \text{re} &= x.\text{re} \otimes d \\ \text{im} &= x.\text{im} \otimes d \\ e &= (EPS/2) \otimes ((1 + EPS) \otimes (|\text{re}| \oplus |\text{im}|)). \end{aligned}$$

PROPOSITION 7.9 (X/D). *If x is an XComplex and d is a double, then $S(x/d) \supseteq S(x)/S(d)$, where*

$$\begin{aligned} x/d &\equiv (\text{re}, \text{im}; e) \text{ with} \\ \text{re} &= x.\text{re} \oslash d \\ \text{im} &= x.\text{im} \oslash d \\ e &= (EPS/2) \otimes ((1 + EPS) \otimes (|\text{re}| \oplus |\text{im}|)). \end{aligned}$$

PROPOSITION 7.10 ($X \times X$). *If x and y are XComplexes, then $S(x \times y) \supseteq S(x) \times S(y)$, where*

$$\begin{aligned} x \times y &\equiv (\text{re}, \text{im}; e) \text{ with} \\ \text{re} &= \text{re1} \ominus \text{re2}, \text{ with } \text{re1} = x.\text{re} \otimes y.\text{re} \text{ and } \text{re2} = x.\text{im} \otimes y.\text{im} \\ \text{im} &= \text{im1} \oplus \text{im2}, \text{ with } \text{im1} = x.\text{re} \otimes y.\text{im} \text{ and } \text{im2} = x.\text{im} \otimes y.\text{re} \\ e &= EPS \otimes ((1 + 2EPS) \otimes ((|\text{re1}| \oplus |\text{re2}|) \oplus (|\text{im1}| \oplus |\text{im2}|))). \end{aligned}$$

Proof. The error is bounded by the sum of the contributions from the real part and the imaginary part:

$$\begin{aligned} &|(x.\text{re} \times y.\text{re} - x.\text{im} \times y.\text{im}) - ((x.\text{re} \otimes y.\text{re}) \ominus (x.\text{im} \otimes y.\text{im}))| \\ &+ |(x.\text{re} \times y.\text{im} + x.\text{im} \times y.\text{re}) - ((x.\text{re} \otimes y.\text{re}) \oplus (x.\text{im} \otimes y.\text{im}))|. \end{aligned}$$

We want to bound this by a machine formula. Let us begin by bounding

$$|(x.\text{re} \times y.\text{re} - x.\text{im} \times y.\text{im}) - ((x.\text{re} \otimes y.\text{re}) \ominus (x.\text{im} \otimes y.\text{im}))|$$

by a machine formula:

$$\begin{aligned}
& |(x.\text{re} \times y.\text{re} - x.\text{im} \times y.\text{im}) - ((x.\text{re} \otimes y.\text{re}) \ominus (x.\text{im} \otimes y.\text{im}))| \\
& \leq |((x.\text{re} \times y.\text{re}) - (x.\text{im} \times y.\text{im})) - ((x.\text{re} \otimes y.\text{re}) - (x.\text{im} \otimes y.\text{im}))| \\
& \quad + |((x.\text{re} \otimes y.\text{re}) - (x.\text{im} \otimes y.\text{im})) - ((x.\text{re} \otimes y.\text{re}) \ominus (x.\text{im} \otimes y.\text{im}))| \\
& \leq |(x.\text{re} \times y.\text{re}) - (x.\text{re} \otimes y.\text{re})| + |(x.\text{im} \times y.\text{im}) - (x.\text{im} \otimes y.\text{im})| \\
& \quad + (EPS/2)|(x.\text{re} \otimes y.\text{re}) - (x.\text{im} \otimes y.\text{im})| \\
& \leq (EPS/2)|(x.\text{re} \otimes y.\text{re})| + (EPS/2)|(x.\text{im} \otimes y.\text{im})| \\
& \quad + (EPS/2)(|x.\text{re} \otimes y.\text{re}| + |x.\text{im} \otimes y.\text{im}|) \\
& = (EPS/2)(2)(|x.\text{re} \otimes y.\text{re}| + |x.\text{im} \otimes y.\text{im}|) \\
& \leq EPS(1 + EPS/2)(|x.\text{re} \otimes y.\text{re}| \oplus |x.\text{im} \otimes y.\text{im}|).
\end{aligned}$$

Almost the exact same calculation produces the analogous formula for the imaginary contribution, and we now combine the two to get a bound on the total error.

$$\begin{aligned}
& \leq EPS(1 + EPS/2)(|x.\text{re} \otimes y.\text{re}| \oplus |x.\text{im} \otimes y.\text{im}|) \\
& \quad + EPS(1 + EPS/2)(|x.\text{re} \otimes y.\text{im}| \oplus |x.\text{im} \otimes y.\text{re}|) \\
& \leq EPS \otimes ((1 + 2EPS) \otimes (|x.\text{re} \otimes y.\text{re}| \oplus |x.\text{im} \otimes y.\text{im}|) \\
& \quad \oplus (|x.\text{re} \otimes y.\text{im}| \oplus |x.\text{im} \otimes y.\text{re}|)). \quad \square
\end{aligned}$$

PROPOSITION 7.11 (D/X). *If x is a double and y is an XComplex, then $S(x/y) \supseteq S(x)/S(y)$, where*

$$\begin{aligned}
x/y & \equiv (\text{re}, \text{im}; e) \text{ with} \\
\text{re} & = (x \otimes y.\text{re}) \otimes nrm \text{ where } nrm = y.\text{re} \otimes y.\text{re} \oplus y.\text{im} \otimes y.\text{im} \\
\text{im} & = -(x \otimes y.\text{im}) \otimes nrm \\
e & = (2EPS) \otimes ((1 + 2EPS) \otimes (|\text{re}| \oplus |\text{im}|)).
\end{aligned}$$

Proof. The true version of x/y is equal to

$$(x \times y.\text{re} + i(-x \times y.\text{im})) / ((y.\text{re})^2 + (y.\text{im})^2)$$

and we need to compare this with the machine version to find the error. Further, this error is less than or equal to the sum of the real error and the imaginary error. Thus, we start with the real calculation (as in the statement of the proposition, we use nrm to represent the machine version of $(y.\text{re})^2 + (y.\text{im})^2$).

$$\begin{aligned}
& \left| \frac{x \times y.\text{re}}{(y.\text{re})^2 + (y.\text{im})^2} - ((x \otimes y.\text{re}) \otimes nrm) \right| \\
& \leq \left| (x \otimes y.\text{re}) \otimes nrm - \frac{x \otimes y.\text{re}}{nrm} \right| \\
& \quad + \left| \frac{x \otimes y.\text{re}}{nrm} - \frac{x \times y.\text{re}}{nrm} \right| + \left| \frac{x \times y.\text{re}}{nrm} - \frac{x \times y.\text{re}}{(y.\text{re})^2 + (y.\text{im})^2} \right|.
\end{aligned}$$

Before continuing, let us compare $\frac{1}{nrm}$ and $\frac{1}{(y.re)^2 + (y.im)^2}$ by developing a formula for comparing $\frac{1}{a^2 + b^2}$ and its associated $\frac{1}{nrm}$:

LEMMA 7.2.

$$\left| \frac{1}{nrm} - \frac{1}{a^2 + b^2} \right| \leq (EPS + (EPS/2)^2) \frac{1}{nrm}$$

where $nrm = a \otimes a \oplus b \otimes b$.

Proof. We compute that

$$\left(\frac{1}{1 + EPS/2} \right)^2 \times nrm \leq a^2 + b^2 \leq \left(\frac{1}{1 - EPS/2} \right)^2 \times nrm;$$

hence

$$\frac{1}{nrm} (1 - EPS/2)^2 \leq \frac{1}{a^2 + b^2} \leq \frac{1}{nrm} (1 + EPS/2)^2.$$

It then follows that

$$\begin{aligned} \left| \frac{1}{nrm} - \frac{1}{a^2 + b^2} \right| &\leq \frac{1}{nrm} (1 + EPS/2)^2 - \frac{1}{nrm} \\ &= \frac{1}{nrm} ((1 + EPS/2)^2 - 1) = (EPS + (EPS/2)^2) \frac{1}{nrm}. \end{aligned} \quad \square$$

Getting back to our main calculation (with $nrm = y.re \otimes y.re \oplus y.im \otimes y.im$), we have

$$\begin{aligned} &\left| (x \otimes y.re) \otimes nrm - \frac{x \otimes y.re}{nrm} \right| \\ &\quad + \left| \frac{x \otimes y.re}{nrm} - \frac{x \times y.re}{nrm} \right| + \left| \frac{x \times y.re}{nrm} - \frac{x \times y.re}{(y.re)^2 + (y.im)^2} \right| \\ &\leq (EPS/2) \frac{|x \otimes y.re|}{nrm} + (EPS/2) \frac{|x \otimes y.re|}{nrm} + (EPS + (EPS/2)^2) \frac{|x \times y.re|}{nrm} \\ &= (EPS/2) \left(\frac{1}{nrm} \right) (2|x \otimes y.re| + (2 + EPS/2) \times |x \times y.re|) \\ &\leq (EPS/2) \left(\frac{1}{nrm} \right) (2|x \otimes y.re| + (2 + EPS/2)(1 + EPS/2) \times |x \otimes y.re|) \\ &= (EPS/2) \left(\frac{1}{nrm} \right) (|x \otimes y.re|)(2 + (2 + EPS/2)(1 + EPS/2)) \\ &\leq (EPS/2)(4 + 3EPS/2 + (EPS/2)^2)(|x \otimes y.re|) \left(\frac{1}{nrm} \right) \\ &\leq (EPS/2)(4 + 3EPS/2 + (EPS/2)^2)(1 + EPS/2)(|x \otimes y.re| \otimes nrm) \\ &\leq (2EPS)(1 + 3EPS/8 + (EPS/4)^2)(1 + EPS/2)(|(x \otimes y.re \otimes nrm)|). \end{aligned}$$

We also get the analogous formula for the imaginary contribution for the error, so our total error is bounded by

$$\begin{aligned}
& (2EPS)(1 + 3EPS/8 + (EPS/4)^2)(1 + EPS/2)(|(x \otimes y.\text{re}) \otimes nrm|) \\
& \quad + (|(x \otimes y.\text{im}) \otimes nrm|) \\
& \leq (2EPS)(1 + 3EPS/8 + (EPS/4)^2)(1 + EPS/2)^2 \\
& \quad \cdot (|(x \otimes y.\text{re}) \otimes nrm| \oplus |(x \otimes y.\text{im}) \otimes nrm|) \\
& \leq (2EPS)(1 - EPS/2)(1 + 2EPS) \\
& \quad \cdot (|(x \otimes y.\text{re}) \otimes nrm| \oplus |(x \otimes y.\text{im}) \otimes nrm|) \\
& \leq (2EPS) \otimes ((1 + 2EPS) \otimes (|(x \otimes y.\text{re}) \otimes nrm| \\
& \quad \oplus |(x \otimes y.\text{im}) \otimes nrm|)).
\end{aligned}$$

Here we used the fact that

$$(1 + 3EPS/8 + (EPS/4)^2)(1 + EPS/2)^2 \leq (1 - EPS/2)(1 + 2EPS). \quad \square$$

This should give the flavor of division proofs. As such, we will skip the proofs of X/X and A/A and simply refer to the *Annals* web site.

PROPOSITION 7.12 (X/X). *If x and y are XComplexes, then $S(x/y) \supseteq S(x)/S(y)$, where*

$$\begin{aligned}
x/y & \equiv (\text{re}, \text{im}; e) \text{ with} \\
\text{re} & = (x.\text{re} \otimes y.\text{re} \oplus x.\text{im} \otimes y.\text{im}) \otimes nrm \\
& \quad \text{where } nrm = y.\text{re} \otimes y.\text{re} \oplus y.\text{im} \otimes y.\text{im} \\
\text{im} & = (x.\text{im} \otimes y.\text{re} \oplus x.\text{re} \otimes y.\text{im}) \otimes nrm \\
e & = (5EPS/2) \otimes ((1 + 3EPS) \otimes A) \text{ where} \\
A & = (|x.\text{re} \otimes y.\text{re}| \oplus |x.\text{im} \otimes y.\text{im}|) \oplus (|x.\text{im} \otimes y.\text{re}| \oplus |x.\text{re} \otimes y.\text{im}|) \otimes nrm.
\end{aligned}$$

PROPOSITION 7.13 (A/A). *If x and y are AComplexes with $y.e < 100EPS \otimes |y|$, or, more accurately,*

$$(y.e)^2 < ((10000EPS) \otimes EPS) \otimes nrm$$

then $S(x/y) \supseteq S(x)/S(y)$, where

$$\begin{aligned}
x/y & \equiv (\text{re}, \text{im}; e) \text{ with} \\
\text{re} & = (x.\text{re} \otimes y.\text{re} \oplus x.\text{im} \otimes y.\text{im}) \otimes nrm \\
& \quad \text{where } nrm = y.\text{re} \otimes y.\text{re} \oplus y.\text{im} \otimes y.\text{im} \\
\text{im} & = (x.\text{im} \otimes y.\text{re} \oplus x.\text{re} \otimes y.\text{im}) \otimes nrm \\
e & = (1 + 4EPS) \otimes (((5EPS/2) \otimes A \oplus (1 + 103EPS) \otimes B) \otimes nrm) \text{ where} \\
A & = (|x.\text{re} \otimes y.\text{re}| \oplus |x.\text{im} \otimes y.\text{im}|) \oplus (|x.\text{im} \otimes y.\text{re}| \oplus |x.\text{re} \otimes y.\text{im}|) \\
B & = x.e \otimes (|y.\text{re}| \oplus |y.\text{im}|) \oplus (|x.\text{re}| \oplus |x.\text{im}|) \otimes y.e.
\end{aligned}$$

In our last proposition we will construct the square-root function. As a warm-up, ignoring round-off error, our construction is as follows. If $x = x.\text{re} + ix.\text{im}$ then $\sqrt{x} = s + id$ where $s = \sqrt{(|x.\text{re}| + h(x.\text{re}, x.\text{im}))/2}$ and $d = x.\text{im}/(2s)$ when $x.\text{re} > 0.0$, and $\sqrt{x} = d + is$ otherwise. Thus, we take our (no-round-off) square roots to be in the first and fourth quadrants.

PROPOSITION 7.14 (\sqrt{X}). *If x is an XComplex, then $S(\sqrt{x}) \supseteq \sqrt{S(x)}$ where we let $s_o = \sqrt[3]{(|x.\text{re}| \oplus h_o(x.\text{re}, x.\text{im}))} \otimes 0.5$ and $d_o = (x.\text{im} \otimes s) \otimes 0.5$, and define*

$$\begin{aligned} \sqrt{x} &\equiv (\text{re}, \text{im}; e) \text{ where} \\ \text{re} &= s_o \text{ if } x.\text{re} > 0.0 \text{ and } \text{re} = d_o \text{ otherwise,} \\ \text{im} &= d_o \text{ if } x.\text{re} > 0.0 \text{ and } \text{im} = s_o \text{ otherwise,} \\ e &= EPS \otimes ((1 + 4EPS) \otimes (1.25 \otimes s_o \oplus 1.75 \otimes |d_o|)). \end{aligned}$$

Proof. This will be a little nasty. Let us begin by analyzing e_s , which is the difference between the true calculation of s and the machine calculation of s , that is $e_s = |s - s_o|$. First, we bound s .

$$\begin{aligned} s &= \sqrt{(|x.\text{re}| + h(x.\text{re}, x.\text{im})) * 0.5} \\ &\leq (1 + EPS)^{1/2} \sqrt{(|x.\text{re}| + h_o(x.\text{re}, x.\text{im})) * 0.5} \\ &\leq (1 + EPS)^{1/2} (1 + EPS/2)^{1/2} \sqrt{(|x.\text{re}| \oplus h_o(x.\text{re}, x.\text{im})) * 0.5} \\ &\leq (1 + EPS)^{1/2} (1 + EPS/2)^{1/2} \\ &\quad \cdot (1 + EPS/2) \sqrt[3]{(|x.\text{re}| \oplus h_o(x.\text{re}, x.\text{im})) * 0.5} \\ &= (1 + EPS)^{1/2} (1 + EPS/2)^{3/2} s_o. \end{aligned}$$

By a power series expansion, we see that

$$\begin{aligned} (1 + EPS)^{1/2} (1 + EPS/2)^{3/2} &= \left(1 + \frac{1}{2}EPS - \frac{1}{8}EPS^2 + \dots\right) \\ &\quad + \left(1 + \frac{3}{2}EPS/2 + \frac{3}{8}(EPS/2)^2 + \dots\right) \\ &= \left(1 + \frac{5}{4}EPS + \frac{11}{32}EPS^2 + \dots\right), \end{aligned}$$

so that,

$$s \leq \left(1 + \frac{5}{4}EPS + \frac{11}{32}EPS^2 + \dots\right) s_o.$$

Similarly,

$$s \geq \left(1 - \frac{5}{4}EPS\right) s_o.$$

Thus, we can bound the s error,

$$\begin{aligned} e_s = |s - s_o| &\leq \left(\left(1 + \frac{5}{4}EPS + \frac{11}{32}EPS^2 + \dots \right) - 1 \right) s_o \\ &= \left(\frac{5}{4}EPS + \frac{11}{32}EPS^2 + \dots \right) s_o. \end{aligned}$$

Next, we analyze e_d , which is the absolute value of the difference between the true calculation of d and the machine calculation of d . That is, $e_d = |d - d_o|$.

$$\begin{aligned} e_d &= |x.\text{im}/(2s) - x.\text{im} \oslash (2s_o)| \\ &\leq |x.\text{im} \oslash (2s_o) - x.\text{im}/(2s_o)| + |x.\text{im}/(2s_o) - x.\text{im}/(2s)| \\ &\leq (EPS/2)|x.\text{im}/(2s_o)| + \left| \frac{x.\text{im}}{2} \frac{s - s_o}{ss_o} \right| \\ &\leq (EPS/2)|x.\text{im}/(2s_o)| + \left| \frac{x.\text{im}}{2} \frac{1}{ss_o} ((5/4)EPS + (11/32)EPS^2 + \dots) s_o \right| \\ &\leq (EPS/2)|x.\text{im}/(2s_o)| \\ &\quad + \left| \frac{x.\text{im}}{2} \frac{1}{s_o(1 - (5/4)EPS)} ((5/4)EPS + (11/32)EPS^2 + \dots) \right| \\ &= (EPS/2)|x.\text{im}/(2s_o)| \left(1 + \frac{(5/2) + (11/16)EPS + \dots}{(1 - (5/4)EPS)} \right) \\ &= (EPS/2) \frac{(7/2) + (-9/16)EPS + \dots}{(1 - (5/4)EPS)} |x.\text{im}/(2s_o)| \\ &\leq (EPS/2)(1 + EPS/2) \frac{7/2}{(1 - (5/4)EPS)} |x.\text{im} \oslash (2s_o)| \\ &= (EPS/2)(1 + EPS/2) \frac{7/2}{(1 - (5/4)EPS)} |d_o|. \end{aligned}$$

Finally, we can bound the overall error $e = e_s + e_d$.

$$\begin{aligned} e_s + e_d &\leq \left(\frac{5}{4}EPS + \frac{11}{32}EPS^2 + \dots \right) s_o \\ &\quad + (EPS/2)(1 + EPS/2) \frac{7/2}{(1 - (5/4)EPS)} |d_o| \\ &\leq \left(EPS + \frac{11}{40}EPS^2 + \dots \right) \left(\frac{5}{4}s_o \right) \\ &\quad + EPS(1 + EPS/2) \frac{1}{(1 - (5/4)EPS)} \left| \frac{7}{4}d_o \right| \\ &\leq EPS(1 + EPS/2) \frac{1}{(1 - (5/4)EPS)} \left(\frac{5}{4}s_o \right) \\ &\quad + EPS(1 + EPS/2) \frac{1}{(1 - (5/4)EPS)} \left| \frac{7}{4}d_o \right| \\ &\leq EPS(1 + EPS/2) \frac{1}{(1 - (5/4)EPS)} \left(\frac{5}{4}s_o + \left| \frac{7}{4}d_o \right| \right) \end{aligned}$$

$$\begin{aligned}
&\leq EPS(1 + EPS/2)^3 \frac{1}{(1 - (5/4)EPS)} \left(\frac{5}{4} \otimes s_o \oplus \left| \frac{7}{4} \otimes d_o \right| \right) \\
&\leq EPS(1 - (EPS/2))(1 + 4EPS) \left(\frac{5}{4} \otimes s_o \oplus \left| \frac{7}{4} \otimes d_o \right| \right) \\
&\leq EPS \otimes \left((1 + 4EPS) \otimes \left(\frac{5}{4} \otimes s_o \oplus \left| \frac{7}{4} \otimes d_o \right| \right) \right). \quad \square
\end{aligned}$$

Now, we develop two formulas for the absolute value of an XComplex.

Formula 7.0 (absUB(X)). If x is an XComplex, then there is an upper bound on the absolute value of x as follows:

$$\begin{aligned}
|x| &= h(x.\text{re}, x.\text{im}) \leq (1 + EPS)h_o(x.\text{re}, x.\text{im}) \\
&\leq (1 - EPS/2)(1 + 2EPS)h_o(x.\text{re}, x.\text{im}) \\
&\leq (1 + 2EPS) \otimes h_o(x.\text{re}, x.\text{im}).
\end{aligned}$$

Thus, we define

$$\text{absUB}(x) = (1 + 2EPS) \otimes h_o(x.\text{re}, x.\text{im}).$$

Formula 7.1 (absLB(X)). If x is an XComplex, then we get a lower bound on the absolute value of x as follows.

$$\begin{aligned}
|x| &= h(x.\text{re}, x.\text{im}) \geq (1 - EPS)h_o(x.\text{re}, x.\text{im}) \\
&\geq (1 + EPS/2)(1 - 2EPS)h_o(x.\text{re}, x.\text{im}) \\
&\geq (1 - 2EPS) \otimes h_o(x.\text{re}, x.\text{im}).
\end{aligned}$$

Thus, we define

$$\text{absLB}(x) = (1 - 2EPS) \otimes h_o(x.\text{re}, x.\text{im}).$$

Finally, in several places in the *verify* program we perform a standard operation on a pair of doubles and must take into account round-off error. This is easy if we use Lemma 7.0.

For example, in *inequalityHolds* we want to show that

$$wh \times wh > \text{absUB}(\text{along}),$$

where $wh = \text{absLB}(\text{whirle})$. By Lemma 7.0, we know that

$$(1 - EPS) \otimes (wh \otimes wh) \leq wh \times wh$$

and we simply test that

$$(1 - EPS) \otimes (wh \otimes wh) \geq \text{absUB}(\text{along}).$$

A slightly more complicated version of this occurs in the computer calculation of $\text{pos}[i]$ and $\text{size}[i]$, that is, the center and size of a sub-box. Prior to

multiplication by $\text{scale}[i] = 2^{(5-i)/6}$, the calculations of pos and size are exact. However, multiplication by scale introduces round-off error. For the center of the sub-box we will have the computer use $\text{pos}[i] \otimes \text{scale}[i]$ with the realization that this is not necessarily $\text{pos}[i] \times \text{scale}[i]$. Thus, we have to choose appropriate sizes to ensure that the machine sub-box contains the true sub-box.

Notationally, this is annoying, because we typically use a computer command like $\text{pos}[i] = \text{pos}[i] \otimes \text{scale}[i]$, while in an exposition, we need to avoid that. We will denote the true center of the sub-box by $p[i]$ and the machine center of the sub-box by $p_0[i]$, and the true and machine sizes will be denoted $s[i]$ and $s_0[i]$. We will let $\text{pos}[i]$ and $\text{size}[i]$ be the position and size (true and machine are the same) before multiplication by $\text{scale}[i]$.

Let $p[i] = \text{pos}[i] \times \text{scale}[i]$, $p_0[i] = \text{pos}[i] \otimes \text{scale}[i]$, and $s[i] = \text{size}[i] \times \text{scale}[i]$. We must select $s_0[i]$ so that $p_0[i] + s_0[i] \geq p[i] + s[i]$. (Here, taking + on the left-hand side is correct, because the need for machine calculation there is incorporated at other points in the programs.) So, we must find $s_0[i]$ such that $s_0[i] \geq (p[i] - p_0[i]) + s[i]$.

$$\begin{aligned}
(p[i] - p_0[i]) + s[i] &\leq (EPS/2)|p_0[i]| + \text{size}[i] \times \text{scale}[i] \\
&\leq (EPS/2)|p_0[i]| + (1 + EPS/2)(\text{size}[i] \otimes \text{scale}[i]) \\
&\leq (1 + EPS/2)((EPS/2)|p_0[i]| + (\text{size}[i] \otimes \text{scale}[i])) \\
&\leq (1 + EPS/2)^2((EPS/2)|p_0[i]| \oplus (\text{size}[i] \otimes \text{scale}[i])) \\
&\leq (1 + 2EPS) \otimes ((EPS/2)|p_0[i]| \oplus (\text{size}[i] \otimes \text{scale}[i])).
\end{aligned}$$

Thus we take

$$s_0[i] = (1 + 2EPS) \otimes ((EPS/2)|p_0[i]| \oplus (\text{size}[i] \otimes \text{scale}[i])).$$

This also works to give $p_0[i] - s_0[i] \leq p[i] - s[i]$. \square

6. AffApproxes with round-off error

In Section 6, we saw how to do calculations with AffApproxes. Here, we incorporate round-off error into these calculations.

Conventions 8.1. An AffApprox x is a five-tuple $(x.f; x.f_0, x.f_1, x.f_2; x.\text{err})$ consisting of four complex numbers $(x.f, x.f_0, x.f_1, x.f_2)$ and one real number $x.\text{err}$. In Section 6, the real number was denoted $x.e$, but it seems preferable to use $x.\text{err}$ here. Recall (Definition 6.4) that an AffApprox x represents the set $S(x)$ of functions from $A = \{(z_0, z_1, z_2) \in \mathbf{C}^3 : |z_k| \leq 1 \text{ for } k \in \{0, 1, 2\}\}$ to \mathbf{C} that are $x.\text{err}$ -well-approximated by the affine function $x.f + x.f_0 z_0 + x.f_1 z_1 + x.f_2 z_2$.

Remark 8.2. A review of Definition 7.2 (XComplexes and AComplexes; loosely, exact and approximate complex numbers) might be helpful at this time.

One approach to round-off error for AffApproxes would be to replace the four complex numbers in the definition of AffApprox by four AComplex numbers complete with their round-off errors, and similarly for the one real number. We will not do this because it would necessitate keeping track of five separate round-off-error terms when we do AffApprox calculations.

Instead, we will replace the four complex numbers by four XComplexes and push all the round-off error into the .err term. Thus, the definition of AffApprox remains essentially the same as in Section 6. We note that, in doing an AffApprox calculation, our subsidiary calculations will generally be on XComplex numbers and produce an AComplex number whose .e term will be plucked off and forced into the .err term of the final AffApprox.

Conventions 8.3. i) In what follows, we will use Basic Properties 7.0 and Lemmas 7.0 and 7.1. Also, the propositions in Section 6 will be utilized; as such, the numbering of the propositions is the same in Sections 6 and 8 (for example, Proposition 6.7 corresponds to Proposition 8.7).

ii) Some notational simplifications will be introduced: $\text{dist}(x)$ before Proposition 8.6, ax before Propositions 8.8, 8.9, 8.11 (the middle usage of ax differs slightly from the other two), and ay before Propositions 8.8, 8.9.

iii) We will try to keep our notation fairly consistent with that of the *verify* computer program, and this will produce some mildly peculiar notation. In particular, in the operations pertaining to Propositions 8.2 and beyond, the resultant AffApproxes will be denoted $(r_f.z; r_f_1.z, r_f_2.z, r_f_3.z; r_error)$ where the first four terms are XComplexes and the last term is a double (technically, $r_f.z$ is the XComplex part of the AComplex r_f and similarly for the $r_f_k.z$ terms). One break with the notation of the programs though is that AffApproxes are called (in the programs) ACJ's, which stands for "Approximate Complex 1-Jets."

iv) The propositions that follow include in their statements the definitions of the various operations on AffApproxes (see Remark 6.6iii).

v) We remind the reader (see Section 7.4) that all machine operations act on machine numbers, and that the various variables appearing in the propositions are assumed to be doubles.

—X:

PROPOSITION 8.1. *If x is an AffApprox then $S(-x) = -(S(x))$ where*

$$-x \equiv (-x.f; -x.f_0, -x.f_1, -x.f_2; x.err).$$

$X + Y$: We analyze the addition of the AffApproxes $x = (x.f; x.f_0, x.f_1, x.f_2; x.err)$ and $y = (y.f; y.f_0, y.f_1, y.f_2; y.err)$. To get the first term in $x + y$ we add the XComplex numbers $x.f$ and $y.f$; which produce the AComplex number $r.f = x.f + y.f$ (see Proposition 7.4), and then we pluck off the XComplex part, which we denote $r.f.z$. The round-off error part $r.f.e$ will be foisted into the overall error term $r.error$ for $x + y$. Similarly for the next three terms in $x + y$.

Abstractly, the overall error term $r.error$ comes from adding the round-off error contributions $r.f.e$, $r.f_0.e$, $r.f_1.e$, $r.f_2.e$ and the AffApprox error contributions $x.err$, $y.err$. Of course, we have to produce a machine version.

PROPOSITION 8.2. *If x and y are AffApproxes, then $S(x + y) \supseteq S(x) + S(y)$, where*

$$x + y \equiv (r.f.z; r.f_0.z, r.f_1.z, r.f_2.z; r.error)$$

with

$$\begin{aligned} r.f &= x.f + y.f, \\ r.f_k &= x.f_k + y.f_k, \\ r.error &= (1 + 3EPS) \\ &\quad \otimes ((x.err \oplus y.err) \oplus ((r.f.e \oplus r.f_0.e) \oplus (r.f_1.e \oplus r.f_2.e))). \end{aligned}$$

Proof. The error is given by

$$\begin{aligned} &(x.err + y.err) + ((r.f.e + r.f_0.e) + (r.f_1.e + r.f_2.e)) \\ &\leq (1 + EPS/2)(x.err \oplus y.err) \\ &\quad + (1 + EPS/2)((r.f.e \oplus r.f_0.e) + (r.f_1.e \oplus r.f_2.e)) \\ &\leq (1 + EPS/2)^3((x.err \oplus y.err) \oplus ((r.f.e \oplus r.f_0.e) \oplus (r.f_1.e \oplus r.f_2.e))) \\ &\leq (1 + 3EPS) \otimes ((x.err \oplus y.err) \oplus ((r.f.e \oplus r.f_0.e) \oplus (r.f_1.e \oplus r.f_2.e))). \end{aligned}$$

To get the last line we used Lemma 7.1. \square

$X - Y$:

PROPOSITION 8.3.: *If x and y are AffApproxes, then $S(x - y) \supseteq S(x) - S(y)$, where*

$$x - y \equiv (r.f.z; r.f_0.z, r.f_1.z, r.f_2.z; r.error)$$

with

$$\begin{aligned} r.f &= x.f - y.f, \\ r.f_k &= x.f_k - y.f_k, \\ r.error &= (1 + 3EPS) \\ &\quad \otimes ((x.err \oplus y.err) \oplus ((r.f.e \oplus r.f_0.e) \oplus (r.f_1.e \oplus r.f_2.e))). \end{aligned}$$

$X + D$: Here, we add the AffApprox $x = (x.f; x.f_0, x.f_1, x.f_2; x.err)$ to the double y . The only terms that change are the first and the last.

PROPOSITION 8.4. *If x is an AffApprox and y is a double, then $S(x + y) \supseteq S(x) + S(y)$, where*

$$x + y \equiv (r.f.z; r.f_0.z, r.f_1.z, r.f_2.z; r.error)$$

with

$$\begin{aligned} r.f &= x.f + y, \\ r.f_k &= x.f_k, \\ r.error &= (1 + EPS) \otimes (x.err \oplus r.f.e). \end{aligned}$$

Proof. The error is given by

$$x.err + r.f.e \leq (1 + EPS) \otimes (x.err \oplus r.f.e)$$

by Lemma 7.0. □

$X - D$:

PROPOSITION 8.5. *If x is an AffApprox and y is a double, then $S(x - y) \supseteq S(x) - S(y)$, where*

$$x - y \equiv (r.f.z; r.f_0.z, r.f_1.z, r.f_2.z; r.error)$$

with

$$\begin{aligned} r.f &= x.f - y, \\ r.f_k &= x.f_k, \\ r.error &= (1 + EPS) \otimes (x.err \oplus r.f.e). \end{aligned}$$

$X \times Y$: We multiply the AffApproxes x and y while pushing all error into the .err term.

We will use the functions (see Formulas 7.0 and 7.1, at the end of Section 7) $\text{absUB} = (1 + 2EPS) \otimes \text{hypot}_o(x.re, x.im)$ and $\text{absLB}(x) = (1 - 2EPS) \otimes \text{hypot}_o(x.re, x.im)$.

When x is an AffApprox, we define $\text{dist}(x)$ to be

$$(1 + 2EPS) \otimes (\text{absUB}(x.f_0) \oplus (\text{absUB}(x.f_1) \oplus \text{absUB}(x.f_2))).$$

This is the machine representation of the sum of the absolute values of the linear terms in the AffApprox x (the proof is straightforward).

PROPOSITION 8.6. *If x and y are AffApproxes, then $S(x \times y) \supseteq S(x) \times S(y)$, where*

$$x \times y \equiv (r.f.z; r.f_0.z, r.f_1.z, r.f_2.z; r.error)$$

with

$$\begin{aligned} r_f &= x.f \times y.f, \\ r_f_k &= x.f \times y.f_k + x.f_k \times y.f, \\ r_error &= (1 + 3EPS) \otimes (A \oplus (B \oplus C)), \end{aligned}$$

and

$$\begin{aligned} A &= (\text{dist}(x) \oplus x.\text{err}) \otimes (\text{dist}(y) \oplus y.\text{err}), \\ B &= \text{absUB}(x.f) \otimes y.\text{err} \oplus \text{absUB}(y.f) \otimes x.\text{err}, \\ C &= (r_f.e \oplus r_f_0.e) \oplus (r_f_1.e \oplus r_f_2.e). \end{aligned}$$

Proof. We add the non-round-off error term for $x \times y$ to the various round-off error terms that accumulated.

$$\begin{aligned} &((\text{dist}(x) + x.\text{err}) \times (\text{dist}(y) + y.\text{err})) + ((\text{absUB}(x.f) \times y.\text{err} \\ &\quad + \text{absUB}(y.f) \times x.\text{err}) + (r_f.e + r_f_0.e) + (r_f_1.e + r_f_2.e)) \\ &\leq (1 + EPS/2)^3 [(\text{dist}(x) \oplus x.\text{err}) \otimes (\text{dist}(y) \oplus y.\text{err})] \\ &\quad + (1 + EPS/2)^2 \{(\text{absUB}(x.f) \otimes y.\text{err} \\ &\quad \oplus \text{absUB}(y.f) \otimes x.\text{err}) + ((r_f.e \oplus r_f_0.e) \oplus (r_f_1.e \oplus r_f_2.e))\} \\ &\leq (1 + EPS/2)^3 A + (1 + EPS/2)^3 (B \oplus C) \\ &\leq (1 + 3EPS) \otimes (A \oplus (B \oplus C)). \end{aligned} \quad \square$$

$X \times D$:

PROPOSITION 8.7. *If x is an AffApprox and y is a double, then $S(x \times y) \supseteq S(x) \times S(y)$, where*

$$x \times y = (r_f.z; r_f_0.z, r_f_1.z, r_f_2.z; r_error)$$

with

$$\begin{aligned} r_f &= x.f \times y, \\ r_f_k &= x.f_k \times y, \\ r_error &= (1 + 3EPS) \\ &\quad \otimes ((x.\text{err} \otimes |y|) \oplus ((r_f.e \oplus r_f_0.e) \oplus (r_f_1.e \oplus r_f_2.e))). \end{aligned}$$

X/Y : For convenience, let $ax = \text{absUB}(x.f)$, $ay = \text{absLB}(y.f)$.

PROPOSITION 8.8. *If x and y are AffApproxes with $D > 0$ (see below), then $S(x/y) \supseteq S(x)/S(y)$, where*

$$x/y = (r_f.z; r_f_0.z, r_f_1.z, r_f_2.z; r_error)$$

with

$$\begin{aligned} r_f &= x.f/y.f, \\ r_f_k &= (x.f_k \times y.f - x.f \times y.f_k)/(y.f \times y.f), \\ r.\text{error} &= (1 + 3EPS) \otimes (((1 + 3EPS) \otimes A \ominus (1 - 3EPS) \otimes B) \oplus C), \end{aligned}$$

and

$$\begin{aligned} A &= (ax \oplus (\text{dist}(x) \oplus x.\text{err})) \otimes D, \\ B &= (ax \otimes ay \oplus \text{dist}(x) \otimes ay) \oplus ((\text{dist}(y) \otimes ax) \otimes (ay \otimes ay)), \\ C &= (r_f.e \oplus r_f_0.e) \oplus (r_f_1.e \oplus r_f_2.e), \\ D &= ay \ominus (1 + EPS) \otimes (\text{dist}(y) \oplus y.\text{err}). \end{aligned}$$

Proof. As usual, we add the round-off errors to the old AffApprox error, taking into account round-off error, working on it bit by bit.

$$\begin{aligned} &(ax + \text{dist}(x) + x.\text{err})/(ay - (\text{dist}(y) + y.\text{err})) \\ &\leq (1 + EPS/2)^2 \\ &\quad \times (ax \oplus (\text{dist}(x) \oplus x.\text{err})) / (ay - (1 + EPS) \otimes (\text{dist}(y) \oplus y.\text{err})) \\ &\leq (1 + EPS/2)^2 (ax \oplus (\text{dist}(x) \oplus x.\text{err})) / \left(\left(\frac{1}{1 + EPS/2} \right) \right. \\ &\quad \left. \times (ay \ominus (1 + EPS) \otimes (\text{dist}(y) \oplus y.\text{err})) \right) \\ &\leq (1 + EPS/2)^4 (ax \oplus (\text{dist}(x) \oplus x.\text{err})) \\ &\quad \otimes (ay \ominus (1 + EPS) \otimes (\text{dist}(y) \oplus y.\text{err})) \\ &\leq (1 + 3EPS) \otimes A. \end{aligned}$$

The next term, being subtracted, requires opposite inequalities.

$$\begin{aligned} &(ax/ay + \text{dist}(x)/ay) + \text{dist}(y) \times ax/(ay \times ay) \\ &\geq (1 - EPS/2)(ax \otimes ay + \text{dist}(x) \otimes ay) \\ &\quad + (1 - EPS/2)(\text{dist}(y) \otimes ax) / \left(\frac{1}{1 - EPS/2} \right) (ay \otimes ay) \\ &\geq ((1 - EPS/2)^4 ((ax \otimes ay \oplus \text{dist}(x) \otimes ay) \oplus ((\text{dist}(y) \otimes ax) \otimes (ay \otimes ay)))) \\ &\geq (1 + EPS/2)(1 - 3EPS)(B) \geq (1 - 3EPS) \otimes B. \end{aligned}$$

Finally, we do the round-off terms

$$((r_f.e + r_f_0.e) + (r_f_1.e + r_f_2.e)) \leq (1 + EPS/2)^2 C$$

and we put these three pieces together:

$$\begin{aligned} &(ax + \text{dist}(x) + x.\text{err})/(ay - (\text{dist}(y) + y.\text{err})) \\ &\quad - ((ax/ay + \text{dist}(x)/ay) + \text{dist}(y) \times ax/(ay \times ay)) \end{aligned}$$

$$\begin{aligned}
& +((r_f.e + r_f_0.e) + (r_f_1.e + r_f_2.e)) \\
& \leq (1 + 3EPS) \otimes A - (1 - 3EPS) \otimes B + (1 + EPS/2)^2 C \\
& \leq (1 + EPS/2)^2 ((1 + 3EPS) \otimes A \ominus (1 - 3EPS) \otimes B) + C \\
& \leq (1 + 3EPS) \otimes ((1 + 3EPS) \otimes A \ominus (1 - 3EPS) \otimes B) \oplus C. \quad \square
\end{aligned}$$

D/X : We divide a double x by an AffApprox y . For convenience, let $ax = |x|$, $ay = \text{absLB}(y.f)$. Having done division out in the previous proposition, we will skip the proof of Proposition 8.9. See the *Annals* web site for the proof.

PROPOSITION 8.9. *If x is a double and y is an AffApprox with $D > 0$ (see below), then $S(x/y) \supseteq S(x)/S(y)$, where*

$$x/y \equiv (r_f.z; r_f_0.z, r_f_1.z, r_f_2.z; r_error)$$

with

$$\begin{aligned}
r_f &= x/y.f, \\
r_f_k &= -(x \times y.f_k)/(y.f \times y.f), \\
r_error &= (1 + 3EPS) \otimes ((1 + 2EPS) \otimes (ax \odot D) \ominus (1 - 3EPS) \otimes B) \oplus C, \\
B &= ax \odot ay \oplus (\text{dist}(y) \otimes ax \odot (ay \otimes ay)), \\
C &= (r_f.e \oplus r_f_0.e) \oplus (r_f_1.e \oplus r_f_2.e), \\
D &= ay \ominus (1 + EPS) \otimes (\text{dist}(y) \oplus y.err).
\end{aligned}$$

X/D : We divide an AffApprox x by a nonzero double y (the computer will object if $y = 0$). The proof is easy and so we delete it.

PROPOSITION 8.10. *If x is an AffApprox and y is a nonzero double, then $S(x/y) \supseteq S(x)/S(y)$, where*

$$x/y \equiv (r_f.z; r_f_0.z, r_f_1.z, r_f_2.z; r_error),$$

with

$$\begin{aligned}
r_f &= x.f/y \\
r_f_k &= x.f_k/y \\
r_error &= (1 + 3EPS) \otimes ((x.err \odot |y|) \oplus [(r_f.e \oplus r_f_0.e) \oplus (r_f_1.e \oplus r_f_2.e)]).
\end{aligned}$$

\sqrt{X} : Here, x is an AffApprox and we let $ax = \text{absUB}(x.f)$. There are two cases to consider depending on whether or not $D = ax \ominus (1 + EPS) \otimes (\text{dist}(x) \oplus x.err)$ is or is not greater than zero.

PROPOSITION 8.11a. *If x is an AffApprox and $D = ax \ominus (1 + EPS) \otimes (\text{dist}(x) \oplus x.err)$ is not greater than zero, then $S(\sqrt{x}) \supseteq \sqrt{S(x)}$, where we use the crude overestimate*

$$\sqrt{x} \equiv \left(0; 0, 0, 0; (1 + 2EPS) \otimes \sqrt[3]{(ax \oplus (x \text{ dist} \oplus x.err))} \right).$$

Proof.

$$\begin{aligned}\sqrt{ax + x \text{ dist} + x \text{ err}} &\leq (1 + EPS/2)\sqrt{(ax \oplus (x \text{ dist} \oplus x \text{ err}))} \\ &\leq (1 + 2EPS) \otimes \sqrt[3]{(ax \oplus (x \text{ dist} \oplus x \text{ err}))}. \quad \square\end{aligned}$$

PROPOSITION 8.11b. *If x is an AffApprox and $D = ax \ominus (1 + EPS) \otimes (\text{dist}(x) \oplus x \text{ err})$ is greater than zero, then $S(\sqrt{x}) \supseteq \sqrt{S(x)}$, where*

$$\sqrt{x} \equiv (r_f.z; r_f_0.z, r_f_1.z, r_f_2.z; r_error)$$

with

$$\begin{aligned}r_f &= \sqrt{x.f}, \\ t &= r_f + r_f, \\ r_f_k &= \text{AComplex}(x.f_k.re, x.f_k.im; 0)/t.\end{aligned}$$

(Simply put, $r_f_k = x.f_k/(2\sqrt{x.f})$. The reason we have to fuss to define r_f_k is because $\sqrt{x.f}$ is an AComplex.)

$$\begin{aligned}r_error &= (1 + 3EPS) \\ &\otimes \left\{ (1 + EPS) \otimes \sqrt[3]{ax} \ominus (1 - 3EPS) \otimes [\text{dist}(x) \otimes (2 \times \sqrt[3]{ax}) \oplus \sqrt[3]{D}] \right\} \\ &\oplus ((r_f.e \oplus r_f_0.e) \oplus (r_f_1.e \oplus r_f_2.e)).\end{aligned}$$

Proof. Let us work on the pieces.

$$\sqrt{ax} \leq (1 + EPS) \otimes \sqrt[3]{ax}.$$

Next,

$$\begin{aligned}\text{dist}(x)/(2\sqrt{ax}) + \sqrt{ax - (\text{dist}(x) + x \text{ err})} \\ \geq (1 - EPS/2)^2 \text{dist}(x) \otimes (2\sqrt[3]{ax}) \\ + (1 - EPS/2)^{1/2} \sqrt{ax \ominus (1 + EPS) \otimes (\text{dist}(x) \oplus x \text{ err})} \\ \geq (1 - EPS/2)^3 [\text{dist}(x) \otimes (2\sqrt[3]{ax}) \oplus \sqrt[3]{D}] \\ \geq (1 + EPS/2)(1 - 3EPS) [\text{dist}(x) \otimes (2\sqrt[3]{ax}) \oplus \sqrt[3]{D}] \\ \geq (1 - 3EPS) \otimes [\text{dist}(x) \otimes (2\sqrt[3]{ax}) \oplus \sqrt[3]{D}].\end{aligned}$$

Adding in the usual term, we get as our error bound

$$\begin{aligned}\sqrt{ax} - (\text{dist}(x)/(2\sqrt{ax}) + \sqrt{ax - (\text{dist}(x) + x \text{ err})}) \\ + ((r_f.e + r_f_0.e) + (r_f_1.e + r_f_2.e)) \\ \leq (1 + EPS) \otimes \sqrt[3]{ax} - (1 - 3EPS) \otimes [\text{dist}(x) \otimes (2\sqrt[3]{ax}) \oplus \sqrt[3]{D}] \\ + (1 + EPS/2)^2 ((r_f.e \oplus r_f_0.e) \oplus (r_f_1.e \oplus r_f_2.e))\end{aligned}$$

$$\begin{aligned}
&\leq (1 + EPS/2)^3 \left(\{ (1 + EPS) \otimes \sqrt[3]{ax} \right. \\
&\quad \ominus (1 - 3EPS) \otimes [\text{dist}(x) \otimes (2\sqrt[3]{ax}) \oplus \sqrt[3]{D}] \} \\
&\quad \oplus ((r_{-f}.e \oplus r_{-f_0}.e) \oplus (r_{-f_1}.e \oplus r_{-f_2}.e)) \Big) \\
&\leq (1 + 3EPS) \otimes \left(\{ (1 + EPS) \otimes \sqrt[3]{ax} \right. \\
&\quad \ominus (1 - 3EPS) \otimes [\text{dist}(x) \otimes (2\sqrt[3]{ax}) \oplus \sqrt[3]{D}] \} \\
&\quad \oplus ((r_{-f}.e \oplus r_{-f_0}.e) \oplus (r_{-f_1}.e \oplus r_{-f_2}.e)) \Big).
\end{aligned}$$

□

7. A trail of breadcrumbs about finding killer words

RECYCLE(For convenience, we will generally refer to a “sub-box” simply as a “box.”)

Here we outline some hints which might be useful in transforming a problem statement similar to the one which inspired this effort into what could eventually be published as the proof of a theorem.

It’s easy to construct a process which would in theory eventually lead to a solution; the difficulty is in ensuring an answer within a practical amount of time and effort.

The first steps are in manipulating the problem statement into an efficiently-addressible form. In this category for this proof were:

- i) realization that the tightness of error estimates would govern success or failure.
- ii) an initial leap to the idea of a self-contained proof using an affine interval calculus. DETAIL(incorporate the round-off error into the remainder term. The main reason for this additional complexity is to get more accuracy in our calculation of *AffApproxes*, which allows us to analyze substantially fewer boxes. Further, the individual computations are faster.)

DETAIL(a killerword which eliminates a point $x \in \mathcal{W}_i$ eliminates all of \mathcal{W}_i . The computational object we constructed to carry out these Taylor approximations is called an *AffApprox*. In the parameter space \mathcal{W} , all functions analyzed via Taylor approximations in this way are built up from the operations $+$, $-$, \times , $/$, $\sqrt{}$. We prove combination formulas for these operations, which show how the Taylor approximations (including the remainder term) change when one of these operations is applied to two *AffApproxes*. This is carried out in Chapter 6.)

- iii) Found a way to tighten error bounds by changing the parameterization to avoid using \exp and \log . DETAIL(This problem can be avoided

slickly by exponentiating the preliminary parameter space \mathcal{P} to get the parameter space \mathcal{W} (the definition of \mathcal{W} is given in Definition 1.22). Lemmas 1.24 and 1.25 then demonstrate that while working in \mathcal{W} one need only understand the basic arithmetic operations $+$, $-$, \times , $/$, $\sqrt{}$. The machine implementation of these basic operations is governed by the IEEE standard IEEE-754 (see [IEEE]).)

The next step was to gather some preliminary data which could support a back-of-the-envelope estimate of the projected time to complete the first search. Here, we:

- i) Wrote a first draft of the affine calculus implementing (only) the operations required.
- ii) Wrote a first draft of the word search algorithm.
- iii) Began experiments with random locations, shrinking region size until the word search found a word which eliminated the entire region.
- iv) Divided the total volume of the compact parameter space by the volume : time ratio to get a crude estimate of the time to complete.

Next, the planned algorithm is tested against the estimate, and iteratively sharpened, measuring progress against the logarithm of the ratio of (maximum reasonable time to complete) to (current projected time to complete).

- i) Moved from geometric combinations into easier-to-manipulate word representation.
- ii) Discovered and exploited the fact that the box center calculus was many times faster than entire box calculus.
- iii) Embraced the concept of sharing the results of the word search across several regions.

Eventually, the estimated time to complete became somewhat reasonable; and after a few last exteraneous small improvements, it was time to switch gears, increasing the difficulty of the attempted tasks.

The first algorithm which was intended to have a chance to finish did not much resemble the final algorithm. It employed depth-first search, a seed-fill type algorithm to spread useful killerwords to adjacent regions and thereby to wherever they were needed; and a region subdivision algorithm which was variable-aspect, intended to adapt to the needs of the killerwords in play.

DETAIL(At that stage, it became evident that the search process, as opposed to the subdivision process, was consuming nearly all of the computation time, and so the second version iterated over regions in breadth-first order,

and, once it found a killerword, tried to use that word on all adjacent regions.
)

To test this first algorithm, it was asked to fill in increasingly-large fractions of parameter space. In doing so, we discovered that we'd been mildly lucky with the first random probes: some regions of parameter space are much harder than others. We kept track of these difficult regions, and used them to test potential improvements.

The second algorithm was a complete rewrite of the first. It employed depth-first search, and within a single program run it kept and retried all killerwords used.

The second algorithm proved good enough to attempt a comprehensive search. Within a day or two, however, it quickly emerged that the search was "exploding", greatly exceeding its computational estimates. After temporarily halting the search, we discovered the predominance of "impossible relators" in the log files related to unsolved regions, and adjusted the theory include a new special case to take these out of consideration.

The second algorithm was in fact good enough to unearth Vol3 (which was known to be in parameter space at the time the search began). So far so good! However, the estimated time to finish started growing and growing, until the estimate exceeded our maximum patience for letting the search finish. Time to stop computing and start thinking again.

ORIGINALSTORY(Rather than blindly selecting words in first-in-first-out order, the algorithm can rank the words under consideration based on a heuristic estimate of the likelihood of their being useful (a word is *useful* if it is a prefix of a killerword). We note first that short words tend to be better than long words, as they have fewer steps and less error. Second, we note that words with a large translation distance are given a bad ranking, for two reasons: they will need more generators appended before they get back to the small translation distance which is needed for a contradiction, and computations with those words introduce more error per step than computations with closer words.

This approach was an improvement, but was not finding enough killerwords in the regions around X_3 and X_4 . Further investigation showed that the algorithm was getting stuck on an identity: once it found an identity, it would consider only words which started with that identity, and ignored all of the other words. To fix this problem, a "diversity" heuristic was introduced, to give special consideration to unlikely but unusual words.)

DETAIL(The third stage of the revision process reduced the number of boxes by attempting all found killerwords in a large region (about a thousand boxes) on all boxes in the region. It did not do any searching, since it was provided with a list of killerwords known to work.)

At this point, several tools were thrown at the problem. We plotted the unsolved boxes in the ndview high-dimensional viewer in geomview (thereby discovering the complex conjugate symmetry that halved the size of parameter space); we ran profiles (discovering that complex affine arithmetic is far slower than we expected). we looked at representative error logs (discovering that our heuristics about box size were mostly irrelevant, since 90% of the time the nonlinear error was turning out unbounded). And, eventually, we discovered that our word search algorithm was a pile of junk, effectively no better than a brute-force search.

Fixing this last opportunity for improvement gave us hope that the search could complete, and in time it did. Back to the blackboard, as the search had unearthed the exceptional regions, and it took considerable effort to find the arguments required to work around them.

It was only when we started writing the first drafts of the manuscript that we discovered the last error: A mistake in the error bounds in one of the calculations. Fixing the code and rerunning the verification was an easy matter, but patching the parts of the proof tree that we had mistakenly believed to be solved took several more weeks of computer effort.

ORIGINALSTORY(The final version was created when the bugs in evaluation were brought to light, and the existing killerword tree was found to be insufficient. It used the list of killerwords used for the entire tree, and some statistics about the number of subdivisions required in order for a given word to kill a particular box, and evaluated each word on each box. Whenever a word was evaluated, a kind of triage was used to determine whether that word was likely to kill the box in question, likely to kill any of its n^{th} generation descendants, or unlikely to kill any descendants of the box; the answer to that heuristic either allowed more detailed evaluation (with the error term included), deferred further evaluation until the box had been subdivided n more times, or excluded that word from further consideration on any descendant of the box. With these heuristics, this program wound up evaluating on average about 10 of the roughly 13200 words per box, and was able to construct the tree consisting of the decomposition into sub-boxes with associated conditions/killerwords.)

In the hindsight of these and other experiences, there are several hints that I'd offer to the brave soul who attempts another ambitious computer-assisted proof:

- i) Isolate the proof verification code from the proof finding code. DETAIL(We mention that the bugs, complexity, and frequent changes in the search programs are irrelevant to the accuracy of the verification. In fact, we shielded the verification programs from internal issues related to the searching programs. Given a putative decomposition of the parame-

ter space into sub-boxes with associated conditions/killerwords the program *verify* simply checks whether this decomposition with conditions/killerwords works.) RIGOR(Finally, there is the issue of rigor. The main difficulty in making the plan work rigorously is that, in computer calculations, we need to account for the difference between the result of an operation and its approximate computed value. The accumulated errors can become significant when performing many operations, such as when using a 43-letter word. A substantial portion of this paper is devoted to addressing the issue of rigor. See also Preview 1.35.)

- XX) Don't be afraid to manually intervene while heuristics are being developed. DETAIL(there was considerable human input to tell the search about particularly difficult killerwords, or to tweak its search parameters (length, and weightings in the heuristic).)
- ii) pay close attention to projected time-to-completion and stats which might predict it.
- iii) whenever possible, use geometry to meet in middle
- iv) emulate Ramanujan (attack risk first)
- v) plan on complete rewrite for proof checking, with K.I.S.S. data passing.
- vi) gather + use stats about calculus to learn which heuristics to use for pruning search tree

In addition, for an application of the method within the context of rigorous enumeration of small 3-manifolds:

- i) word library Keep track of all best killerwords (those which are included in the current smallest known killertree), and reuse them wherever possible.
- ii) corner trick When deciding whether or not to use a word, First calculate whether or not the word kills the center of the box and each of its first dimension count ancestors. If in some case it doesn't, word can't kill the box, as each of those centers is in the box. This guard to the much more time-consuming box calculation can be made very fast by combining preorder tree traversal, short-circuit evaluation, and lazy evaluation.
- iii) Use breadth-first search at a high level, and depth-first search at the detail level. breadth-first search allows embarrassingly parallelism, facilitates a top-down understanding of overall progress, and maximizes the effectiveness of the word library. depth-first search facilitates allows exploitation of local concerns, e.g., the corner trick.

- iv) lazy + depth- or breadth- stratified word search DETAIL(To prevent the search from running forever, it is temporarily abandoned after some number of steps, and re-done with twice as many steps every time the number of descendant boxes doubles. This way, the search could run forever, but only if the subdivision process runs forever.)

ENDMARKER

8. Computer code

FIXME

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