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Department of Computer Science



Technical Report

An Implementation of CAD in Maple Utilising McCallum Projection

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An implementation of CAD in Maple utilising McCallum projection

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Abstract

Cylindrical algebraic decomposition (CAD) is an important tool for the investigation of semi-algebraic sets. Originally introduced by Collins in the 1970s for use in quantifier elimination it has since found numerous applications within algebraic geometry and beyond. Following from his original work in 1988, McCallum presented an improved algorithm, CADW, which offered a huge increase in the practical utility of CAD.

In 2009 a team based at the University of Western Ontario presented a new and quite separate algorithm for CAD, which was implemented and included in the computer algebra system MAPLE. As part of a wider project at Bath investigating CAD and its applications, Collins and McCallum's CAD algorithms have been implemented in MAPLE. This report details these implementations and compares them to QEPCAD and the Ontario algorithm.

The implementations were originally undertaken to facilitate research into the connections between the algorithms. However, the ability of the code to guarantee order-invariant output has led to its use in new research on CADs which are minimal for certain problems. In addition, the implementation described here is of interest as the only full implementation of CADW, (since QEPCAD does not currently make use of McCallum's delineating polynomials), and hence can solve problems not admissible to other CAD implementations.

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1 Introduction

Cylindrical algebraic decomposition (CAD) was first announced in 1973 by Collins. A CAD is a decomposition of \mathbb{R}^n into cells, constructed with respect to a set of input polynomials in n variables. Each cell can be described as a semi-algebraic set and the cells are cylindrically arranged, meaning the projection of any two cells is either equal or disjoint. Usually, the CAD produced is such that each polynomial is sign-invariant within each cell, allowing for the solution of many problems defined by the polynomials. Collins provided the first algorithm to compute a CAD [11, 1], developed as a tool for quantifier elimination in real closed fields. Since their discovery they have found numerous applications ranging from robot-motion planning [20] to simplification technology [4, 14, etc.]

Collins' algorithm has two phases. The first, projection, applies a projection operator repeatedly to a set of polynomials, each time producing another set of polynomials in one fewer variables. Together these sets contain the projection polynomials. The second phase, lifting, then builds the CAD incrementally from these polynomials. First \mathbb{R} is decomposed into cells which are points and intervals corresponding to the real roots of the univariate polynomials. Then \mathbb{R}^2 is decomposed by repeating the process over each cell using the bivariate polynomials at a sample point of the cell. The output for each cell consists of sections of polynomials (where a polynomial vanishes) and sectors (the regions between these). Together these form the stack over the cell, and taking the union of these stacks gives the CAD of \mathbb{R}^2 . This process is repeated until a CAD of \mathbb{R}^n is produced.

To conclude that the CAD of \mathbb{R}^n produced using sample points in this way is sign-invariant we need the key concept of delineability. A polynomial is *delineable* in a cell if the portion of its zero set in the cell consists of disjoint sections. A set of polynomials are *delineable* in a cell if each is delineable and further that the sections of different polynomials in the cell are either identical or disjoint. A projection operator is valid for use in the algorithms if over each cell of a sign-invariant CAD for projection polynomials in r variables, the polynomials in r+1 variables are delineable.

The output of a CAD algorithm depends on the ordering of the variables. In this paper we usually work with polynomials in $\mathbb{Z}[x_1,\ldots,x_n]$ with the variables, \mathbf{x} , listed in ascending order; (so we first

project with respect to x_n and so on until we have univariate polynomials in x_1). The main variable of a polynomial, mvar(f), is the greatest variable present with respect to the ordering.

Since Collins published the original algorithm there has been much research into improvements. These include, but are not restricted to; the use of equational constraints [19, etc.], the notion of partial CADs [13, etc.] and [21, etc.], ideas on cell adjacency and clustering [2, etc.], ideas on preprocessing the input [23, etc] and heuristics to pick the best variable ordering [15, etc.]. A summary of developments over the first twenty years of CAD was given by [12].

A key area of research has been in improving the projection operator used in the first phase of the algorithm. Collins original operator can produce CADs with many more cells than required for sign invariance of the polynomials. McCallum proved in [17] that for most problems a far simpler projection operator could be used. Then in [18] McCallum presented an algorithm, named CADW, detailing how and when the improved operator should be used. A particular feature of the algorithm is that the CAD produced is not just sign invariant with respect to the input polynomials, but order invariant. Further development of McCallum's algorithm and operator have been presented by Brown in [6, 8].

McCallum's projection operator is implemented in QEPCAD [7]; an interactive command-line program written in C using the SACLIB library of computer algebra functions. The name stands for *quantifier elimination by partial cylindrical algebraic decomposition* but the software can also produce full cads and offers a wide range of options such as the use of equational constraints and 2d plots.

In 2009 a new approach to producing CADs was presented in [10]. Instead of projection and lifting this approach first computes a decomposition of complex space and then refines to real space, making use of the theory of triangular sets and regular chains [3, etc.]. The algorithm was implemented in MAPLE and is included in the RegularChains library [16, etc.], distributed with MAPLE.

This report gives details on a new implementation of CAD via projection and lifting, within MAPLE. The implementation was originally undertaken to facilitate research into the connections between the two approaches for computing CAD. However, its ability to guarantee order-invariant output has led the implementation to utility as part of new algorithm in [5] offering decompositions which while not sign invariant for their input polynomials, are minimal for certain problems based on them. In addition, the implementation is of interest as it is the only full implementation of CADW, since QEPCAD can not produce McCallum's delineating polynomials leading to unnecessary warnings about potential failure for certain examples. (We note that SyNRAC [24] also implements CAD via projection and lifting in MAPLE, but this implementation uses Collins' algorithm and so can not provide order-invariance for use in [5].)

In Section 2 we summarize the theory of the Collins and McCallum projection operators algorithm and in Section 3 we describe our implementation in Maple. In Section 4 we describe how this implementation compares with both Qepcad and the algorithm from the theory of regular chains. This report should be accompanied with a file ProjectionCAD.mm containing the code implementing the algorithms in Maple. The file is designed to be read into Maple to define a library of commands, but can also be viewed as a text file. Details on installing and working with the code are given in Subsection 3.1.

2 Projection operators

We summarize the projection operators of Collins and McCallum, (for more details see [1] and [17]).

Let f and g be real univariate polynomials. Denote by $\deg(f)$ the degree of f, by f' the first derivative of f, by $\operatorname{red}^k(f)$ the kth reductum of f, by $\operatorname{discr}(f)$ the discriminant of f, by $\operatorname{res}(f,g)$ the resultant of f and g and by $\operatorname{psc}_i(f,g)$ the jth principal subresultant coefficients of f and g.

Let F be a set of univariate polynomials and suppose $f,g\in F$. Then denote by $\operatorname{coeff}(F)$, the set of all non-zero coefficients of all elements of F, by $\operatorname{red}(F)$ the set of all $\operatorname{red}^k(f)$ such that $0\leq k\leq \deg(f)$, by $\operatorname{psc}(F)$ the set of all $\operatorname{psc}_j(f,g)$ such that $f\neq g$ and $0\leq j\leq \min(\operatorname{def}(f),\operatorname{deg}(g))$, by $\operatorname{psd}(F)$ the set of all $\operatorname{psc}_j(f,f')$ such that $0\leq j\leq \deg(f)-1$, by $\operatorname{discr}(F)$ the set of all $\operatorname{discr}(f)$ and by $\operatorname{res}(F)$ the set of $\operatorname{res}(f,g)$. We can now define the operators

```
\begin{array}{lcl} PROJ(F) &:= & \mathrm{coeff}(F) \cup \mathrm{psd}(\mathrm{red}(F)) \cup \mathrm{psc}(\mathrm{red}(F)), \\ P(F) &:= & \mathrm{coeff}(F) \cup \mathrm{discr}(F) \cup \mathrm{res}(F). \end{array}
```

Note that P(F) is a subset of PROJ(F). When dealing with multivariate polynomials we may think of them as univariate polynomials in the main variable and use the definitions as above.

When Collins first devised CAD he showed that P may be used as the projection operator for problems in 2 variables. However, P cannot always be used for problems in an arbitrary number of variables and so [11] specified the algorithm using PROJ instead. In [17] McCallum proved that P may also be used

for problems in 3 variables and then in [18] McCallum extended his results to explain when they hold for problems in more variables.

Key to McCallum's proofs is the concept of order invariance. Recall that a polynomial has order of vanishing k at a point, if k is the smallest non-negative integer such that some partial derivative of the polynomial of order k does not vanish at the point. A CAD is **order invariant** with respect to a set of polynomials if each polynomial has constant order of vanishing within each cell.

Let F be a square free basis of polynomials in r variables. McCallum's key theorem showed that if P(F) is order invariant over a cell of \mathbb{R}^{r-1} then each element of F either vanishes identically or is order invariant in every cell of the stack above. Hence the theorem can be applied recursively to prove P is valid, so long as the nullification of a polynomial over a lower-dimensional cell does not occur.

To classify when the operator P may be used, McCallum defined the following condition. Let F be a set of polynomials in r variables and let prim(F) and cont(F) be respectively the set of primitive parts and the set of contents of F. Then F is **well-oriented** if either r = 1 or both the following hold:

- (1) Every $f \in \text{prim}(F)$ has a finite number of nullification points, i.e. if the main variable is x_k then $f(\alpha, x_k) = 0$ for at most a finite number of $\alpha \in \mathbb{R}^{k-1}$.
- (2) The set $P(F) \cup \text{cont}(F)$ is well oriented.

If condition (1) is changed to the stronger condition of *no nullification points* then the polynomials are said to be **very well oriented**.

For well-oriented sets of polynomials the projection operator P can be used to generate CADs. This is shown using McCallum's theorem discussed above and, in the case where are a finite number of nullifying points, by replacing the nullified polynomial f with a **delineating polynomial**; a partial derivative of f whose roots include the x_k coordinate of every point where the order of f differs from the minimal order induced by α . In [8] Brown defined the **minimal delineating polynomial** as the case where these are the only roots.

In [18] McCallum gave an algorithm CADW which aims to build a CAD using P. Before lifting over each cell the algorithm checks for nullification over the cell. If the dimension of the cell is zero then a delineating polynomial is used instead and otherwise the algorithm fails, declaring the polynomials not well-oriented. If the algorithm finishes then the CAD produced is order-invariant.

QEPCAD offers both McCallum's operator and Hong's modification of Collins' [13]. However, QEPCAD does not implement the calculation of delineating polynomials and so can fail unnecessarily when using McCallum's operator in cases where the polynomials are well oriented but not very well-oriented. ¹

However, this situation is minimised since before declaring failure QEPCAD ensures this is really necessary by running through some checks detailed in [8]. For example, it may be the case that nullification does not lead to order-invariance, (meaning the minimal delineating polynomial is a constant). Consider f=zy-x which is nullified by the point (x,y)=(0,0) and so is well oriented but not very well oriented. However, for all values of z this polynomial has order 1 and so no delineating polynomial need be added, (and QEPCAD will not fail). Other checks performed by QEPCAD include; seeing if a delineating polynomial is already included in the projection set, checking whether the nullified polynomial is one that must actually be order-invariant to ascertain correctness, and checking whether it is the final lift in which case only sign-invariance is required.

These checks rule out many cases where we may expect QEPCAD to fail but the example constructed below demonstrates that there are still cases where delineating polynomials are really required to guarantee correct output using McCallum's operator.

Example 2.1. To construct the example we need a polynomial in k variables nullified by a cell in k-1 variables. We must have k > 2 or the nullification will be avoided when considering the content and primitive part individually. A simple choice is $f = zy - x^2$ which is nullified over a cell with x = y = 0, which would certainly be produced from its coefficients belonging to the projection set. Further, this f has non-constant minimal delineating polynomial, z.

However, if we were to just construct a CAD for f then the nullification would occur at the final lift and so we need not bother with a delineating polynomial. We must instead consider a polynomial p which produces f as a projection factor. If we let f be a coefficient then we need only make f order-invariant [6] so instead we choose a polynomial for which f is the discriminant: $p = f + w^2$.

Hence, when considering a CAD for p with respect to the variable ordering w > z > y > x, a delineating polynomial will certainly be required. This example was considered in the experiments detailed in Section 4 and using McCallum's operator a CAD with 73 cells was produced.

 $^{^{1}}$ Actually QEPCAD does not necessarily fail, but it will produce a warning message indicating that the output may not be correct.

3 Implementation in Maple

Algorithms to construct CADs with both Collins' and McCallum's projection operators have been implemented in Maple, with pseudo code for the main algorithms presented below and the actual code freely available from the author's website. The package is titled ProjectionCAD and is designed to complement the alternative approach of constructing CADs via regular chains which is already distributed with Maple. Indeed, the code described here makes use of tools from the RegularChains package and gives output in the same format.

Algorithm 1 computes a set of projection polynomials (including the input polynomials) with respect to the chosen operator. Algorithm 2 will use those polynomials and the projection choice to build a CAD of \mathbb{R}^n over which they are sign invariant. Algorithm 3 is a composition of the two which can generate a variety of CADs via projection and lifting.

```
Algorithm 1: CADProjection

Input : • The input set of polynomials F \subset \mathbb{R}[x_n, \dots, x_1].
• A choice proj of either Collins or McCallum.

Output: • A set of projection polynomials P \subset \mathbb{R}[x_n, \dots, x_1].

1 Set P_1 to be the finest square-free basis for the primitive parts of F.
2 Set cont to be the set of contents of F.
3 for i = 2, \dots, n do
4 | Set \hat{P}_i to be proj(P_i) \cup cont excluding any constant polynomials.
5 | Set P_i to be the finest square-free basis for the primitive parts of \hat{P}_i.
6 | Reset cont to be the set of contents of \hat{P}_i.
7 P := \bigcup_{i=1}^n P_i
8 return P
```

Both Collins and McCallum note in their work that for some examples simplifications could be made to the coefficients / reducta included in their operators. For example, if a polynomial has a constant coefficient then this clearly does not need to be added to the projection polynomials. Further, no subsequent polynomials would need to be added since there will be no situations where these become leading coefficient. Some such elementary simplifications have been incorporated in the MAPLE implementation of Algorithm 1.

We note that Algorithm 2 differs depending on the choice of projection operator in Algorithm 1, due to the requirements to check that the input polynomials are well-oriented when using McCallum's operator. There is a further optional input in Algorithm 2 which can be used to request that the CAD outputted is not only sign-invariant but also order invariant. If not set to true then the algorithm does not check for nullification on the final lift since without this the CAD of \mathbb{R}^n is still guaranteed to be sign-invariant, sufficient for most applications, (but insufficient for the application in [5]).

Algorithm 4 is a sub-algorithm which describes how the stack generation is implemented. To generate a stack the real roots of those projection polynomials with main variable x_i need to be calculated when the other variables are set according to a cell of dimension i-1. New cells of dimension i are then defined with the variable x_i set in turn to be those roots and the intervals between. The roots may not be rational but algebraic numbers, and so care needs to be taken in implementing this. We make use of the RegularChains library and in particular and internal command for stack generation described in Section 5.2 of [10]. Algorithm 4 ensures that the polynomials passed to the RegularChains algorithm satisfy the assumptions that algorithm makes, namely that the polynomials are co-prime and square-free when evaluated on the cell, (separate above the cell in the language of regular chains).

3.1 Working with the ProjectionCAD library in Maple

This report should be accompanied by a file ProjectionCAD.mm containing the code implementing the algorithms in Maple. The file is designed to be read into Maple to define a library of commands, but can also be viewed as a text file. To read the package into Maple and make the commands available use

```
> read("ProjectionCAD.mm"):
> with(ProjectionCAD):
```

Commands implementing Algorithms 1-3 are now available, with the same names. For each there are two required arguments; a list of polynomials and a list of variables in descending order. There are also

Algorithm 2: CADLifting

```
Input : • A set of polynomials P \subset \mathbb{R}[x_n, \dots, x_1] from CADProjection.
              • A choice proj of either Collins or McCallum.
              \bullet A boolean fcad if the final CAD is to be order invariant.
   Output: A sign-invariant CAD of \mathbb{R}^n, also order-invariant if fcad = true.
 1 for i = 1, ..., n do
 2 | P_i := \{ p \in P \text{ such that the main variable of p is } x_i \}.
 3 Set C_1 to be a CAD of \mathbb{R} formed by the decomposition of the real line according to the real roots
   of the polynomials in P_n.
 4 for i = 2, ..., n do
       for each cell c \in C_{i-1} do
 5
           if proj = McCallum then
 6
 7
              if i < n or fcad = true then
 8
                  set Q_i to be the empty set.
                  for each polynomial p \in P_{n+1-i} do
 9
                      if p = 0 throughout c then
10
11
                          if dim(c)=0 then
                              add the minimal delineating polynomial to Q_i if it is non-constant.
12
13
14
                              give warning message about potential failure.
15
                       else
                          add p to Q_i
16
               else
17
                | Set Q_i := P_i
18
           else
19
            Set Q_i = P_i
20
           \mathtt{Set}\ S_c := \mathtt{CADGenerateStack}(c,Q_i)
21
       Set C_i := \bigcup_c S_c
22
23 return C_n
```

Algorithm 3: CADFull

```
Input : • The input set of polynomials F \subset \mathbb{R}[x_1, \dots, x_n].

• A choice proj of either Collins or McCallum.

• A boolean fcad if the final CAD needs to be order invariant.

Output: C, a CAD of \mathbb{R}^n.

1 P := \text{CADProjection}(F, proj)
2 C := \text{CADLift}(P, proj, fcad)
3 return C
```

Algorithm 4: CADGenerateStack

```
Input: • A cell c from a CAD of \mathbb{R}^{i-1}, or \emptyset if i=0.
• A set of projection polynomials P \subset \mathbb{R}[x_{n-i},\ldots,x_n].

Output: \mathcal{S}, a stack over c with respect to P.

1 Set \hat{P} to be a set of polynomials with the same zeros as P but which are coprime and square-free within the cell c.

// By encoding the variables which are set to a point as a regular chain, and using commands from the RegularChains package [16].

2 S := \text{RegularChains-GenerateStack}(c, \hat{P})

3 return S
```

optional keyword arguments; method to choose between Collins or McCallum, output to specify how the CAD is represented and final0I to specify that the outputted CAD be order invariant. Details on the progress of the algorithm can be accessed by setting the infolevel for each command.

Every cell in a cad of \mathbb{R}^n is equipped with at least an index and a sample point. The index is an n-tuple of integers indicating the value or range of each variable using the real roots in increasing order computed for each stack. (Hence a cell whose index consists of only even integers defines a single point of \mathbb{R}^n while an index of only odd integers is a cell of full dimension n.) The sample points produced by ProjectionCAD are encoded as algebraic numbers given by regular chains and bounds isolating a single root. This is a consequence of using the internal RegularChains algorithm and is desired as it makes comparing CADs produced by the two approaches simple.

The implementation in MAPLE can also produce cell representations (bounds or values for the variables using algebraic numbers) and if requested (with the output argument) can display the output intuitively using the piecewise construct in MAPLE. The number of cells in the CAD may be measured using the nops command on the output, (or the CADNumCellsInPiecewise command if the piecewise output format is used).

A simple example of using the code is given below. The output is as displayed in a MAPLE worksheet, except that the sample points have been replaced by SP for brevity.

```
> f := x^2+y^2-1:
```

> cad := CADFull([f], vars, method=McCallum, output=piecewise);

$$\begin{cases} SP & y < 0 \\ SP & y = 0 \\ SP & 0 < y \end{cases}$$

$$\begin{cases} SP & y = -\sqrt{-x^2 + 1} \\ SP & y = -\sqrt{-x^2 + 1} \\ SP & x = -\sqrt{-x^2 + 1} \end{cases}$$

$$\begin{cases} SP & x = -1 \\ SP & x = -1 \end{cases}$$

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> CADNumCellsInPiecewise(cad);

13

4 Experimental results

We have run experiments comparing three of the implementations of CAD.

All the tests were run on a Linux desktop with Intel core i5 CPU (1.6GHz) and 8.0Gb total memory. The timings in Maple were recorded using the inbuilt time function while the timings for Qepcad are the total system time recorded at the end of each session. For the tests in Maple we use the timelimit function to limit each example to 1000 seconds^2 of computation time. For the tests in Qepcad we run with the option +N500000000 + L200000, where the first option specifies the memory to be pre-allocated and the second the number of prime numbers to be used.

The examples for the experiments were all taken from the CAD example repository described in [22] and stored at http://opus.bath.ac.uk/29503 and . Some of the problems here are quantified, however these experiments are designed to compare the CAD implementations only, and so were run on the unquantified versions, (i.e. full sign-invariant CADs were computed for the polynomials involved in the problems).

The three implementations compared were as follows.

²The Maple timelimit of 1000s does not apply to kernel operations so it is possible to have timings that are slightly higher, as is the case for the simplified Putnam example in Table 1.

- **PCAD** We let PCAD denote the implementation of CAD via projection and lifting which is the topic of this paper. Run in MAPLE 16 through the command CADFull from the ProjectionCAD package developed at Bath. We run with the default settings; using McCallum projection and worrying about nullification at all except the final lift where this is ignored.
- TCAD We let TCAD denote the implementation of CAD via triangular sets. Run in MAPLE 16 with the in-built command from the RegularChains package, CylindricalAlgebraicDecompose.
- QCAD We let QCAD denote the implementation of CAD via projection and lifting given by QEPCAD-B version 1.69. The code is given the full-cad option and default settings otherwise; McCallum projection and lifting with various improvements [6, 8] but no delineating polynomials.

We note that there are other implementations of CAD available, including Mathematica, Redlog and SyNRAC.

The results of the experiments are displayed in Table 1. An F indicates failure for theoretical reasons ³ while a T/O indicates failure due to timeout in MAPLE. We note that with the setting described above, QEPCAD never failed due to timeout or memory issues.

Unless it fails for theoretical reasons, QCAD is usually the quickest. The examples where it was slower were generally small examples in which the time for QCAD was dominated by its initialisation. It is likely that QEPCAD still represents the state of the art for computing full CADs at the moment. However, we make the cautionary note that the algorithm for TCAD described in [10] is to be replaced by a significantly more efficient version, designed to reduce repeated computations and perform some steps in an incremental manner. This improvement is described in the preprint [9], but the implementation is still under development and not yet available.

Next we note that TCAD never fails for theoretical reasons and is usually (but not always) quicker than PCAD. Comparing PCAD with QCAD we see that the cell counts are usually identical, as expected since they implement the same theoretical algorithm. There are occasions when QCAD will fail when PCAD does not because it does not implement McCallum's delineating polynomials, as in the example from Section 2 and the Quartic example. However, there are also examples where PCAD failed when QCAD did not, (the Whitney umbrella and the x-axis ellipse problem). These theoretical failures were avoided in QCAD since they correspond to cases where nullification does not contradict the theory underpinning the algorithm, which QEPCAD checks for (as discussed in Section 2). It is possible to instruct ProjectionCAD to continue the computation following a failure, which for these examples would lead to the same number of cells as QCAD. The other examples where the cell count differs are due to QCAD using a partial implementation of the simplified operator described in [6].

5 Summary

We have described an implementation of CAD via projection and lifting in MAPLE. The implementation offers both McCallum's and Collins' operators with a variety of customizations available. The implementation was undertaken to facilitate research into the connections between the different approaches to CAD. The experimental results demonstrate that for most problems QEPCAD is superior. Nevertheless, the approach described here offers significant utility for two main reasons.

- (1) It implements McCallum's delineating polynomials (actually Brown's minimal delineating polynomial) and thus can produce the only full implementation of McCallum's CADW. We have demonstrated that there are examples in which QEPCAD will fail for which this implementation can succeed.
- (2) It can offer the user the choice of an order-invariant CAD in the final output. This is essential for the application in [5] which defines a new type of CAD, minimal for certain problems.

Examples have been given where the output of CAD via Regular Chains in MAPLE is not order-invariant and it is not currently known how the algorithm could be modified to produce order-invariance, (further research is currently being undertaken).

It is likely that a small tweak order the hood in QEPCAD could be made to offer order-invariant output at in (2). However, this would then greatly amplify the utility described in (1) since the majority of cases of nullification occur at the final lift where they can be safely ignored if only a sign-invariant CAD is required. For example, consider again $f = zy - x^2$ from Section 2. We noted there that f has non-constant minimal delineating polynomial and so to give an order-invariant CAD of f alone, this would need to be included. Running PCAD gives a CAD of 21 cells (as does TCAD and QCAD). One of

³this includes the production of warnings that the output may not be correct

these cells has x = 0, y = 0, x free, over which f is not order-invariant. However, if we run CADFull using the optional argument to specify an order-invariant output then this cell is split and the CAD outputted has 23 cells.

If this implementation remains essential for applications such as [5] then some future improvements to the package are likely. These should include the implementation of some of the improvements described in [8] to avoid failure unless it is absolutely necessary. Further, an implementation of the Brown-McCallum operator [6] is very desirable to offer the lowest possible cell counts.

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Problem	PCAD		TCAD		QCAD	
	cells	$_{ m time}$	cells	$_{ m time}$	cells	$_{ m time}$
Parametric parabola	115	0.5	27	0.1	115	1.6
Whitney umbrella	F	-	895	3.4	895	3.8
Quartic	333	3.9	233	2.7	F	-
Sphere & catastrophe	509	6.2	421	4.140	509	3.7
Arnon-84	55	0.3	55	0.157	55	3.7
Arnon-84-2	41	0.4	41	0.332	41	3.7
Implicitization	F	-	895	5.228	F	-
Ball & cylinder	365	4.2	365	3.914	365	3.8
Term rewrite system	1099	8.2	1099	7.768	1099	3.6
Collins and Johnson	3673	50.2	3673	65.438	3673	3.8
Lower bounds range	F	-	333	1.550	F	-
X-axis ellipse problem	F	-	20225	252.0	62645	5.1
Davenport & Heintz	4949	23.8	4949	21.0	4949	3.8
Hong-90	27	0.1	27	0.2	27	3.5
Solotareff-3	-	T/O	_	T/O	243325	12.3
Collision problem	-	T/O	-	T/O	45979	5.0
Random trivariate	-	T/O	-	T/O	877	14.9
Off-center ellipse	4569	153.9	2705	51.5	4593	5.0
Concentric circles	41	0.2	41	0.2	41	3.7
Non-concentric circles	41	0.2	41	0.3	41	3.7
Simplified ESP	-	T/O	_	T/O	56105	10.2
Simplified Putnum	55021	1026.0	10517	193.0	10517	3.9
Simplified YangXia	-	T/O	_	T/O	6313	5.2
Simplified SEIT	-	T/O	_	T/O	F	-
Cyclic-3	381	3.7	381	4.2	381	3.8
Example from §2	73	0.2	67	0.1	F	-

Table 1: Table detailing timings and cell counts for computation of full CADs using the three implementations described in Section 4.