

The 2013 Evaluation of SMT-COMP and SMT-LIB

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Abstract After 8 years of SMT Competitions, the SMT Steering Committee decided, for 2013, to sponsor an evaluation of the status of SMT benchmarks and solvers, rather than another competition. This report summarizes the results of the evaluation, conducted by the authors. The key observations are that (1) the competition results are quite sensitive to randomness and (2) the most significant need for the future is assessment and improvement of benchmarks in the light of SMT applications. The evaluation also measured competitiveness of solvers, general coverage of solvers, logics, and benchmarks, and degree of repeatability of measurements and competitions.

1 Introduction

1.1 The Competition history and goals

From 2005 through 2012 (and in 2014), the SMT community sponsored an annual competition among SMT solvers (cf. Fig. 1). The purpose of the competition is to encourage advances in SMT solver implementations acting on benchmark formulas of theoretical or practical interest. Public competitions are a well-known means of stimulating advancement in software tools. For example, in automated reasoning, the SAT and CASC competitions for propositional and first-order reasoning tools, respectively, have spurred significant innovation in their fields [5,6]. Indeed, the SMT competition increased in size each year: more benchmarks were added, new

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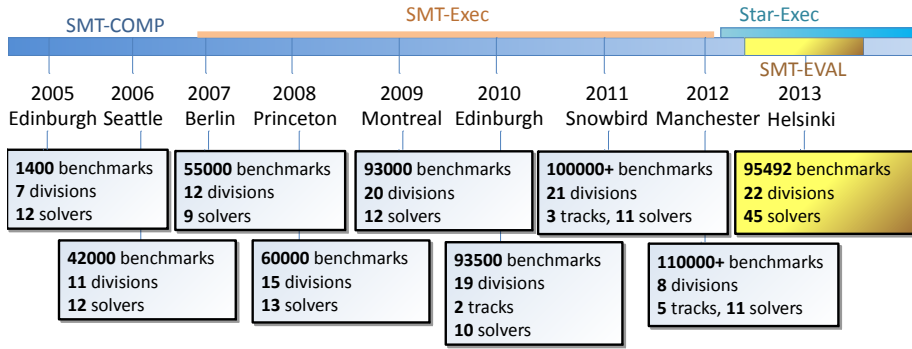


Fig. 1: SMT-COMP history

solver teams participated, solver performance improved, and more logic divisions were defined.

The particular goals of SMT-COMP include the following [10]:

- enable research on SMT solvers by benchmarking and comparing performance;
- promote a standard format for SMT problems (SMT-LIB v2 [4]);
- collect additional benchmarks;
- identify and develop new theories and logics for SMT, encouraging their inclusion in SMT solvers;
- introduce SMT users and implementors to each other;
- provide a forum for SMT implementors to promote their SMT solvers and for SMT users to assess the comparative performance of solvers; and
- encourage the development of industrial-strength solvers for wide-spread use.

1.2 Concerns prompting an evaluation

In 2013, the SMT Steering Committee decided to take a collective breath and sponsor an evaluation of the current state of the art, without the pressure of a competition. In particular, the implementation teams found that preparing for a competition required considerable engineering work that detracted from other goals. On the one hand, the engineering work is necessary for users to use the tools as off-the-shelf applications. But holding the competition every year was causing some otherwise highly involved teams to withdraw. Thus there was a collective desire to pause the competition cycle for a year.

Second, there was a desire to evaluate the state of the SMT community and the competition. The competition had become focused on details of performance on a particular set of benchmarks for particular logics and there was a feeling that its overall results had become somewhat predictable. An evaluation of progress of the SMT community was desired, with a consideration of other metrics and goals, not just winning narrowly focused competitions.

And third, for many years the competition had relied on the SMT-Exec computational infrastructure [3]. A new infrastructure, StarExec [7, 11] had been funded by the NSF and developed at the University of Iowa, but had not yet been tried out in earnest. In fact, a goal in 2012 had been to use StarExec for SMT-COMP 2012; however, StarExec was not sufficiently operational and the competition had to revert to SMT-Exec. It was anticipated that an evaluation in 2013, with less deadline pressure, would enable using StarExec on a shake-down cruise prior to a competition in 2014 at the Federated Logic Conference (FLoC) Olympic Games [9].

1.3 Evaluation topics

The SMT Steering Committee appointed a team of evaluators (the authors of this report) to examine areas of interest. The evaluators, who are not affiliated with any solver team, independently studied the following topics:

- continuity and turnover in SMT solver participation in past competitions (3.1)
- the performance of all historical and current solvers on the full set of benchmarks, measuring
 - the improvement in performance over time (3.4)
 - repeatability of performance measurements (3.5.1)
 - repeatability of competition results (3.5)
 - competitiveness of solvers (3.6)
 - parallel vs. sequential performance (3.7)
- the usefulness of various logics
 - characteristics of existing logics (3.2)
 - which logics are implemented by solvers (3.3)
 - which logics are particularly relevant to application areas (3.9)
- the state of existing benchmarks
 - the range of computational difficulty of the benchmarks (3.8)
 - the degree to which benchmarks discriminate among solvers (3.8)
 - which logics have support in benchmarks (3.8)

2 Evaluation tools

2.1 SMT-LIB benchmarks

One goal of the SMT project since its inception has been collecting benchmarks by which to evaluate SMT solvers and to represent challenge problems in the field. The growth in the number of benchmarks available is shown in the graphics of Fig. 1. All¹ of the benchmarks present in StarExec at the time of evaluation (95 492 benchmarks) were used for the evaluation, including those with unknown status. Of course, in previous years only a subset of these were available, since new benchmarks were added from year to year. Also in 2012, some benchmarks were

¹ One benchmark, added late, was not included in all of the data accumulation; hence some tables report 95 491 benchmarks and some 95 492.

used in the competition that were not yet added into the SMT-LIB benchmark set and were not yet in StarExec at the time of the evaluation. The 2012 benchmark set also included some incremental benchmarks that were not used in this evaluation. Furthermore, the competition each year used a random selection of benchmarks (guided by a difficulty distribution). The fact that the set of benchmarks used in competitions was different each year muddled any year-to-year comparison of results. By using all available benchmarks in the evaluation, we ensure that the current assessment is not susceptible to random choices in benchmark selection (cf. Section 3.5). The various kinds of benchmarks and their distribution across logics are discussed below in Section 3.

2.2 Solvers

The SMT competitions required that participating solvers be publicly available Linux applications, and that they be available for any future experimenter to use on new experiments. Thus all historical solvers are still available. However, in 2010, the competition adopted the then new benchmark format, SMT-LIB v2 [4]. Thus solvers prior to 2010 do not run on the current benchmark set. The participating solvers are shown in Fig. 2.

SMT-EVAL used all historical solvers since 2010 (32 total), added 9 versions of previous solvers that were updated in 2013, and included 4 additional experimental solvers,² for a total of 45 solvers. All solver implementation teams that we could reach were apprised of the upcoming evaluation and given the opportunity to submit new versions of their solvers. Some teams simply submitted the current version of their solver or advised us to download the current public version from the team’s website. Thus the solvers are not necessarily tuned to particular application domains or for competition. Any comparisons for particular applications or kinds of benchmark problems should perform an independent analysis.

2.3 Interpreting solver output

Given the large number of solver-benchmark combinations, it was not feasible to manually inspect the output of each job pair. Thus, we needed to determine mechanically whether a solver reported a benchmark as satisfiable, unsatisfiable or unknown (or did not return a result within the time limit).

On StarExec, this is done by providing a *post-processor*, i.e., an executable that operates on the output of each job pair. Unfortunately, two features of StarExec made it difficult to interpret solver output in a way that is fully SMT-LIB v2 compliant. First, StarExec collects the entire output of a job pair in a single file before passing it to the post-processor. Thus, it is not possible to reliably determine which output a solver has generated specifically in response to a benchmark’s `check-sat` command. Second, StarExec conflates standard output and standard

² These were four variations of a portfolio-style solver submitted by Abziz [1,2].

Solver	Affiliation	2005	2006	2007	2008	2009	2010	2011	2012	2013
Number participating		11	11	9	13	12	10	13	13	9
Abziz...	Cairo U.							••	3+	
Boolector	JKU				X	X		+	+	•
CVC/CVCLite/CVC3	NYU	X	X	X	X	X	+	+	+	
CVC4	NYU						+	+	+	•
MathSat-HeavyBV	Trento								+	
MathSAT 3,4,5	FBK	X	X	X	X	X	+	+	+	•
SMTInterpol	U. Freiburg							+	+	•
SONOLAR	U. Bremen						+	+	+	•
STP(2), simplifyingSTP	Stanford U., MIT		X			X	+	+	+	
4Simp	U. Melbourne								+	
Tiffany de Wintermonte	U. Melbourne								+	
opensmt	U. Lugano				X	X	+	+		•
veriT	UFRN					X	+	+		•
Z3	MSR			X	X			+		•
AProVE NIA	RWTH Aachen						+	+		
MiniSMT	U. Innsbruck						+			•
test_pmathsat	FBK-IRST						+			
barcelogic	UPC	X	X	X	X	X				
beaver	UC Berkeley				X	X				
clsat	Washington U.				X	X				
Sateen	U. Col-Boulder	X	X	X	X	X				
Spear				X	X					
sword	U. Bremen				X	X				
Yices	SRI	X	X	X	X	X				
alt-ergo					X					
ArgoLib				X						
Fx7				X						
Ario		X	X							
ExtSat			X							
HTP		X	X							
Jat			X							
NuSMV			X							
Sammy		X								
SBT		X								
Simplics		X								
SVC		X								

Fig. 2: Solvers used in each year of SMT-COMP and SMT-EVAL. Solvers prior to 2010 (marked by X) do not support SMT-LIB v2 and were not used for SMT-EVAL. Boxes marked with a • symbol identify solvers that are new for the evaluation; all solvers since 2010 (marked with + or • symbols) were used for the evaluation.

error. We hope that this behavior will be changed in the future, but for SMT-EVAL this made it impossible to distinguish between regular solver responses and error messages.

To determine the result of each job pair, we used a post-processor that searched for the words `sat`, `unsat` and `unknown` in the output file, and reported the corresponding result if it found exactly one of them; otherwise, it reported no result. This conservative approach may have caused a small number of (otherwise valid) solver responses to be discarded. It was adopted to minimize the number of incorrect results reported by the post-processor.

2.4 Benchmark scrambling

SMT-COMPs in previous years used benchmark scrambling [3]. We did not scramble benchmarks for SMT-EVAL, mainly because support for this feature was not yet available in StarExec. In principle, this means that solvers could have cheated by matching benchmark contents or filenames against a database of known SMT-LIB benchmarks, or even by simply extracting the `:status` information present in most benchmarks. Ruling out cheating with certainty would require careful inspection of solver sources, which are not available for all solvers. However, based on the evaluation data, and taking into account the lack of strong incentives for cheating in SMT-EVAL, we have no reason to believe that such shortcuts to success occurred.

An additional use of benchmark scrambling is that the change in syntax and identifiers can, for some solvers, change the search paths or the preprocessing that is performed. Thus scrambling serves to exercise different code paths and adds an additional dimension of testing and performance measurement to competitions and evaluations. It would be interesting to investigate these effects of scrambling in detail; we did not do this as part of SMT-EVAL.

2.5 StarExec

SMT-EVAL successfully used the new StarExec computational framework [7, 11]. Running SMT-EVAL on StarExec did indeed expose a number of bugs and user interface issues; these were corrected in the course of the SMT-EVAL runs. Thus SMT-EVAL served a valuable purpose in preparing StarExec for larger scale use.

SMT-EVAL’s largest computational job was running all 45 of the evaluated solvers on all of the relevant non-incremental benchmarks in the SMT library. The benchmarks belong to different *logics* and solvers are characterized by which logics they support. So for each logic, the evaluation executed the cross product of all benchmarks for that logic and all solvers (in all years) that support problems in that logic, for a total of 1 663 478 solver-benchmark combinations (called *job-pairs* in StarExec). This job took several months of wall-clock time to run. We ran it in quarters, using the result of the first quarter to adjust our procedures and debug some of StarExec, before running the rest of the solver-benchmark job pairs. Because of a bug (now corrected), the results of about 600 job-pairs were not present in the accumulated results. These were identified and rerun as an additional “mop-up” job. Each solver-benchmark pair was run in an independent environment, on a computation node by itself, with time and memory limits, as described in Section 3.7.

3 Evaluation results

The evaluation team’s observations on the questions it considered are presented in the following subsections; our overall conclusions are listed in Section 4. The raw

year	2005	2006	2007	2008	2009	2010	2011	2012	2013
# of participants	11	11	9	13	12	10	11	11	9
# dropping out		4	6	2	3	7	2	4	
# new participants	(11)	4	4	6	2	5	3	4	

Fig. 3: Turnover in solver team participation. There are no entries for changes in 2013 because in that year there was no competition with entrants.

data used as the basis for our observations, collected from StarExec, are all archived on the SMT-COMP website, at <http://sourceforge.net/p/smtcomp/code/HEAD/tree/trunk/smtcomp-web/2013/data> as 7z-compressed files. (The largest is 25 MB.) The data can also be viewed directly on StarExec, at <https://www.starexec.org/starexec/secure/explore/spaces.jsp?id=4566>, using its guest login.

3.1 Solver participation

The historically participating solvers are shown in Fig. 2. There are a number of observations to be made about solver participation.

- As shown in Fig. 3, the number of solvers participating each year has been fairly constant, ranging from 9-13, with a median value of 11 participants. The data for 2013 is an anomaly in two respects. The count of 9 only includes those solvers that were explicitly submitted new for 2013. Second, Abziz contributed a portfolio solver that makes an automatic choice, based on machine learning, of which among other existing solvers to apply to a benchmark using observed features of the benchmark. Abziz submitted two instances of his portfolio that used solvers from 2011; those instances of his portfolio solver are not counted in Fig. 3 since they were not available in 2011, though they are included in Fig. 2. In addition, three variations were contributed in 2012; we count just one participation unit in Fig. 3 for 2012, though all three (and the two 2011 entries) were used in the evaluations described in this paper.
- Though any team dropping out of future years’ competitions is to be regretted, such turnover is to be expected: research projects and Ph.D. students move on to other topics. The year of the most drop-outs is 2010, when the benchmark format changed; not all teams could invest the effort to change their front-ends and to accommodate the new semantic features of SMT-LIB v2.
- Accompanying the drop-outs is a roughly equal number of new participants each year; roughly 1/3 of the participants each year are new. One of the goals of an organized competition is to foster new participation, providing a venue in which it is easy to participate and self-evaluate against the state-of-the-art. Our observation is that this goal is being accomplished, despite the significant effort required to implement a reasonably competitive solver.
- Despite the turnover, some teams have participated regularly throughout the history of the competition: CVC3/4 and MathSat have participated since the beginning; several others have participated in most of the years since they began being involved.

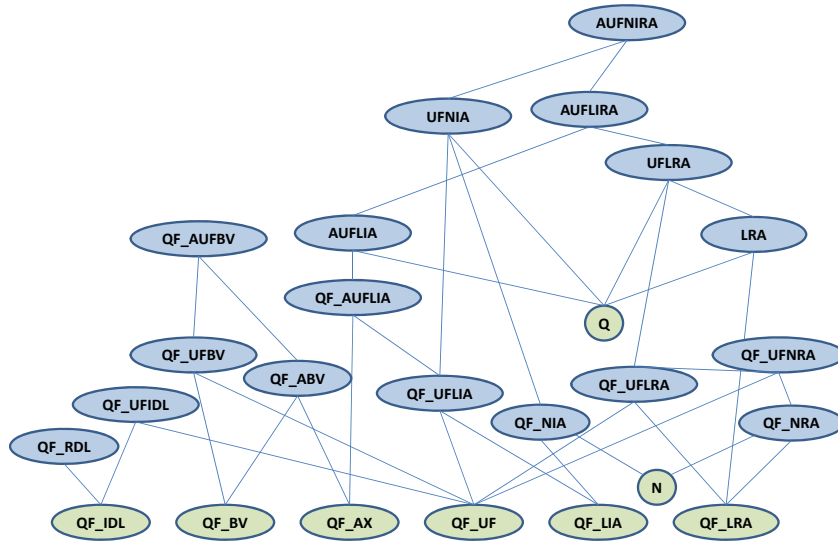


Fig. 4: Graph of SMT logics defined in SMT-LIB. The links represent a natural inclusion by characteristics, but *not* a sub-logic relationship. QF_ABV is defined, but is often subsumed in QF_AUFBV, and is not usually listed separately in other tables in this report. Q is not a logic by itself but is a characteristic indicating that quantified expressions are permitted in the logic; similarly N indicates non-linear arithmetic is allowed.

3.2 Logics

SMT-LIB defines a number of theories and logics. The logics are a combination of theories with additional functions or logic symbols or restrictions on the vocabulary of the theories. For example, the QF_IDL logic uses the underlying Ints theory, but restricts terms to equalities and inequalities between constants and simple differences of variables. The combination of the Ints theory and the Reals theory adds functions that convert between integers and reals (among other things).

Leaving aside difference logics as special cases of integer and real logics, there are essentially these mostly orthogonal characteristics underlying the theories and logics:

- IA: integers
- RA: reals
- N: non-linear arithmetic
- A or AX: arrays

- UF: uninterpreted functions
- QF_: quantifier freeness
- BV: bit-vectors

This list of characteristics partially corresponds to the set of SMT-LIB theories. The exceptions are quantification and non-linearity, both of which designate lifting restrictions on the kinds of terms allowed in the resulting logic. (Since they do not correspond exactly to underlying SMT theories, the term *characteristics* is used in this discussion.)

It is possible to define a logic with any powerset of these characteristics in an almost purely combinatorial sense. Barring non-linearity without arithmetic and the useless empty set, there are 111 combinations of characteristics (instead of the full 128). SMT-LIB has definitions and benchmarks for 20 of these logics (excluding QF_IDL, QF_RDL, and QF_UFIDL) at the time of the evaluation, expanded to 34 in 2014. Fig. 4 shows what might be expected to be natural containment relationships among the logics defined at the time of this evaluation.

However, what might be expected to be natural containment relationships are not actually sub-logic relationships. That more accurate relationship is shown in a diagram on the SMT-LIB website: <http://smtlib.cs.uiowa.edu/logics.shtml> (due to Cesare Tinelli). The principal cause of this lack of sub-logic relationship is that many of the logics supporting arrays support only specific sorts of arrays. For example, the basic QF_AX logic allows arbitrary new sort symbols, including arrays over those free sorts. However, QF_ABV (arrays and bit-vectors) does not allow new sort symbols and only allows arrays over bit-vector sorts.

The naming convention for logics is almost but not quite simply a combination of the letters indicated in the list above, corresponding to the characteristics of the logic, with the letters listed in order by convention:

$$[QF_][A|AX][UF][BV][N|L][IA|RA|IRA].$$

The non-uniformities are these:

- quantification is indicated by *removing* a QF_ prefix;
- the absence of non-linearity (N), i.e., linearity, is indicated by an L, instead of by no designator;
- integer and real arithmetic are indicated by two letters (IA and RA) and their combination by IRA;
- the A used in the arithmetic designators could be ambiguous with the designator for arrays, except for position;
- a logic with just arrays is named QF_AX, whereas in other combinations including arrays, just an A instead of AX is used.

For example, UFNIA includes quantification (no QF_ prefix), uninterpreted functions (UF), non-linearity (N), and integer arithmetic (IA), leaving out arrays, bit-vectors, and reals.

These names have roots in the history of development of SMT-LIB. Nevertheless as the set of logics grows—23 here and 34 by the 2014 competition—the situation will increasingly confuse new users of SMT-LIB trying to understand the structure of logics; in addition new theories and accompanying logics are anticipated. It would

solvers (as stated by the solver teams) is shown in Fig. 5. CVC and Z3 support all or nearly all logics, with CVC4 just missing full support for non-linear arithmetic (it does have partial support, with full support in the 2014 competition); veriT and MathSat support a significant number of logics.

The numbers in the bottom row of the figure state the number of current (2013) solvers that support a given logic. The characteristics most lacking are quantifiers and nonlinearity. Those with the most support are bitvectors, uninterpreted functions and arithmetic.

Note that the support indicated is that stated by the solver supplier at the time the solver was submitted to a competition. In some cases, a solver supports one logic but is not listed as supporting a subset of that logic. The general reason is that the subset, with restricted characteristics, allows for specific optimizations that are not implemented. The more general solver could act on problems of the more restricted benchmarks but is not deemed competitive and thus was not formally entered into the competition for the restricted logic.

3.4 Progress in solver performance

We can obtain a measure of the overall year-by-year improvement in solver performance by applying each year's solvers to all the benchmarks (within each logic), observing the best performance on each benchmark. That is, the data shows a *virtual best performer*, obtained by an all-knowing oracle choosing, for each benchmark, the solver that performs best on that benchmark.

Fig. 6 shows the fraction of benchmarks completed (y-axis) within a given time (in seconds, on the x-axis); thus points toward the upper left are better (more completed in less time). The results for all logics are shown together. The four curves are the data for the four years from 2010 to 2013. The lowest (marked with fine dots) curve is that for 2010. There is clearly significant progress made from 2010 to later years: the number of benchmarks completed within a given time is noticeably higher; we did not attempt to identify the source of this historical improvement. The curves for 2011 and 2012 (heavy dashes and short dashes, respectively) are clearly above 2010, but are not substantially different from each other. The solvers for 2013 (heavy solid line) are uniformly above those for previous years, but not by a large amount.

By this data, the improvements in raw SMT solver performance on the current set of benchmark problems have slowed down, though they may have improved by other measures.

An alternate measure of solver progress is to count, for a given logic, what fraction of the benchmarks have a better time in a given year than in the previous year, by the best solver for each year. That data is shown in Fig. 7. Though most data points indicate that most benchmarks improved year to year, the progress is not uniform. The progress from 2010 to 2013 shows nearly all logics having improvement rates above 80%, but there are definitely some laggards. There is not necessarily a monotonic improvement because the set of solvers is different

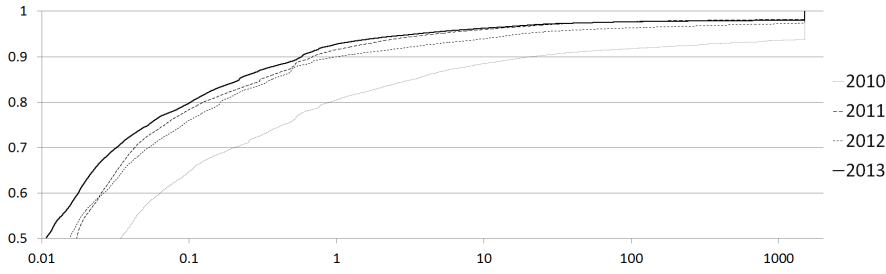


Fig. 6: The fraction of benchmarks (y-axis) whose winning time is less than the given number of seconds (x-axis), by year.

from year to year; some well-performing solvers may choose not to participate every year.

3.5 Repeatability of competitions

The SMT competition aims to be a repeatable measurement of relative performance of solvers. To that end the details of the competition, such as the rules and selection of benchmarks, are made public. In addition the solvers themselves are required to be made public after the competition.

However, there are aspects of the competition that are not deterministic. We measured two of those in this evaluation:

- *The repeatability of the time a given solver takes on a given benchmark* (Section 3.5.1). This variation is a combination of two factors: any variation in execution and timing by the StarExec system itself, and any non-determinism in the execution of the solver.
- *Variation caused by selection of benchmarks* (Section 3.5.2).

There is the additional risk that the set of benchmarks in SMT-LIB is not representative of any particular application area. We do not evaluate that risk here, but expect that this is quite likely the case. As noted in our conclusions, we consider assessment of benchmarks as the most important aspect of future work. It would consist of a survey of the uses of SMT solvers, determination of the corresponding characteristics needed of solvers, an assessment of the degree to which current benchmarks are representative of the identified uses, and collection of additional benchmarks.

3.5.1 Accuracy of performance measurements

To measure the repeatability of a given solver on a given benchmark, we produced a random selection of benchmark-solver pairs (job-pairs), selecting only among job-pairs that did not timeout and had a running time of at least 0.1 seconds. The random selection produced 10 165 job-pairs (out of the total 1 663 478 job-pairs).

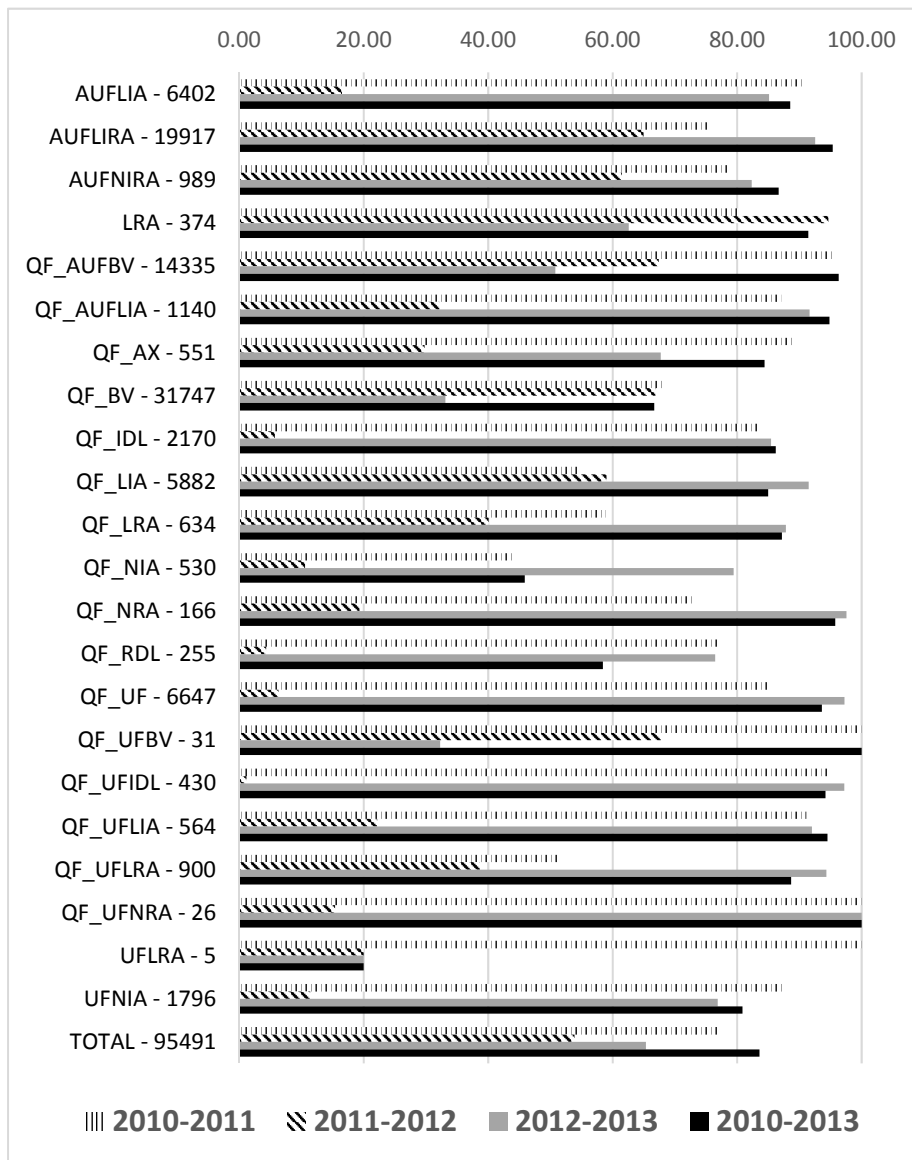


Fig. 7: The fraction of benchmarks whose best time improved between the given years.

These job-pairs were run 8 times on 8 different days. Each such run took only a few hours. This resulted in 8 repetitions of each of the selected job-pairs. The data set is archived with the other evaluation data on the SMT-EVAL website. StarExec was lightly loaded on these days, so we did not measure any effect owing to interference with other uses of StarExec.

The 8 measured values for a job-pair were sorted and the first, median, and third quartile measured; these are as the average of sorted values 2 and 3, 4 and 5, and 6 and 7, respectively, and designated as Q1, Q2, and Q3. We noted the following points:

- Of the measured job-pairs, about 78% had an interquartile range (difference between Q3 and Q1) less than 10% of the median value. That is, repeated measurements will fall within about 10% of each other roughly 50% of the time.³
- 3.5% of job-pairs showed an interquartile range more than 50% of the median value.
- Nine cases of the 10165 had a Q3 value more than double the median. These clearly showed multi-modal behavior among the eight data points for each benchmark-solver pair. Such behavior is likely the result of non-deterministic choices within the solver’s search algorithms.
- We did not expand our study to determine whether the variation in performance was correlated to the choice of solver or to certain benchmarks. Also, we did not measure any effects owing to benchmark scrambling, which we now know can change solver behavior significantly.

An approximate conclusion from the above is that variation in timing itself plays a relatively small role in any variation in the competition. We expect the variation averages out over many benchmarks and is not correlated with particular solvers. It might have a larger effect on benchmarks with short running times (but we did not quantify this). The number of benchmarks per division used in previous competitions was on the order of 200, but did vary from division to division; the standard deviation of a mean calculated from a sample of 200 observations of an approximately normal distribution would be only about 0.05 of the interquartile range.

Any variation due to non-determinism within a solver we consider part of that solver’s design. Thus it is possible that in some competition a non-deterministic solver might do quite well, while in a repeat of the same competition that solver, making other internal random choices, might do poorly.

3.5.2 Accuracy of competitions

The data we collected allows us to simulate various competition organizations. One such competition design would be, for each logic, to run all solvers for the given year on all benchmarks. The result of such an exercise is shown in Figs. 8 and 9 for the set of 9 solvers for the year 2013. By summing all the CPU times for each

³ We did not measure skew: the degree to which Q3 and Q1 are unequally distant from the median.

benchmark-solver combination, we obtain that such a competition would use about 443 days of wall-clock time (using one computational node); for this computation we consider all timeout and memory exhaustion results to have taken the full timeout value (1500 seconds) of time. It is possible that in some cases a solver may exhaust memory prior to the timeout.

In previous competitions, the winner was determined as the solver that correctly solved the most benchmarks within the timeout limit (solvers producing incorrect results are disqualified). Ties among solvers solving the same number of benchmarks are resolved in favor of the solver taking the least amount of time to produce its solutions. In the tables showing the results, for each logic, the results are reported in the order of a virtual winner.

However, past competitions have not used all benchmarks, but rather a random subset of a selection of the benchmarks. It is worth asking how susceptible the competition results are to the particular subset of benchmarks used for the competition. Typically, the choice was not completely random; rather, the distribution was (a) constrained to have roughly equal representation of various categories of difficulty and (b) curated to avoid over-representation of benchmark families that contained many similar benchmarks. For our evaluation here, we determined the results of a virtual competition by simply randomly selecting different equal-sized subsets (ignoring difficulty measures), determining the competition results for each, and observing whether the competition results vary significantly depending on the subset. Fig. 10 shows the result of an experiment in which a virtual competition was executed on 1000 random subsets; we tallied the fraction of times that the winner was different than the winner in the full virtual competition (Figs. 8–9) and the fraction of times that the complete order of the finishers was different. (We limited our consideration to those logics with at least 100 benchmarks and with more than one solver.)

Note that we did not curate the benchmarks to address the issue of over-representation of some benchmark families (the authors of this report are not knowledgeable about which families are at issue); we simply used all of the benchmarks. Thus the study should not be seen as a competition that ranks solvers, but as one that investigates various competition organizations.

The results are quite instructive. In many logics, the competition winners change in only a small fraction of trials. However, in several logics, the fraction of trials in which the winner changes is quite significant, in some cases over 60%. Inspection of the data reveals two contributing causes.

- First, some logics have a relatively large number of benchmarks that are unsolved by some solver within the timeout limit. The fraction of benchmarks that are not solved is shown in the final column of Fig. 10 (this is the average fraction over all the solvers for that logic). Random selection of benchmarks can readily change the subset of unsolved benchmarks included in the virtual competition or change the balance of benchmarks solved by one solver vs. another. Since “winning” the virtual competition is determined primarily by the number of benchmarks solved, rather than the time, any change in the relative number of solved benchmarks may change the order of winners.

# unsolved	secs	solver
AUFLIA 6402 benchmarks , 71.53 days wall time		
936	1404424.66	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
1278	1957553.10	CVC4-SMT-EVAL-2013
1873	2818137.76	veriT-SMT-EVAL-2013
AUFLIRA 19917 benchmarks , 27.95 days wall time		
254	389439.43	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
638	976484.46	CVC4-SMT-EVAL-2013
680	1049109.48	veriT-SMT-EVAL-2013
AUFNIRA 989 benchmarks , 0.14 days wall time		
8	12012.02	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
LRA 374 benchmarks , 3.49 days wall time		
68	105139.09	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
127	196132.91	CVC4-SMT-EVAL-2013
QF_AUFBV 14335 benchmarks , 64.28 days wall time		
520	805442.63	Boolector-1.5.118-SMT-EVAL-2013
543	844767.74	SONOLAR-2013-05-15-SMT-EVAL-2013
712	1106772.52	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
864	1332157.54	MathSAT5-5.2.6-SMT-EVAL-2013
948	1465081.78	CVC4-SMT-EVAL-2013
QF_AUFLIA 1140 benchmarks , 3.40 days wall time		
16	24087.55	veriT-SMT-EVAL-2013
23	34635.67	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
72	113550.39	CVC4-SMT-EVAL-2013
81	121754.16	MathSAT5-5.2.6-SMT-EVAL-2013
QF_AX 551 benchmarks , 1.39 days wall time		
19	28520.99	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
23	34530.42	MathSAT5-5.2.6-SMT-EVAL-2013
38	57088.31	CVC4-SMT-EVAL-2013
QF_BV 31747 benchmarks , 127.04 days wall time		
944	1491616.08	Boolector-1.5.118-SMT-EVAL-2013
1102	1750286.98	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
1469	2471860.33	MathSAT5-5.2.6-SMT-EVAL-2013
1590	2498512.90	SONOLAR-2013-05-15-SMT-EVAL-2013
1699	2763975.38	CVC4-SMT-EVAL-2013
QF_IDL 2170 benchmarks , 27.81 days wall time		
231	407530.76	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
360	689225.35	CVC4-SMT-EVAL-2013
751	1305966.10	veriT-SMT-EVAL-2013
QF_LIA 5882 benchmarks , 86.65 days wall time		
172	340070.98	MathSAT5-5.2.6-SMT-EVAL-2013
316	614640.90	SMTInterpol-2.0r8402-SMT-EVAL-2013
366	646587.14	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
761	1482588.14	CVC4-SMT-EVAL-2013
2910	4403063.03	veriT-SMT-EVAL-2013
QF_LRA 634 benchmarks , 2.96 days wall time		
7	15398.96	CVC4-SMT-EVAL-2013
19	33604.65	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
23	43613.79	MathSAT5-5.2.6-SMT-EVAL-2013
36	80377.24	veriT-SMT-EVAL-2013
37	82658.86	SMTInterpol-2.0r8402-SMT-EVAL-2013
QF_NIA 530 benchmarks , 3.26 days wall time		
17	28080.56	MiniSMT-0.5-SMT-EVAL-2013
160	253992.63	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013

Fig. 8: (Part 1) Results of a virtual competition using the 2013 solvers on all benchmarks.

# unsolved	secs	solver
QF_NRA 166 benchmarks , 1.64 days wall time		
0	1.64	veriT-SMT-EVAL-2013
12	22751.00	MiniSMT-0.5-SMT-EVAL-2013
79	118521.63	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
QF_RDL 255 benchmarks , 2.57 days wall time		
39	64875.77	CVC4-SMT-EVAL-2013
39	65459.76	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
51	91419.17	veriT-SMT-EVAL-2013
QF_UF 6647 benchmarks , 12.56 days wall time		
30	46692.19	veriT-SMT-EVAL-2013
62	106340.93	CVC4-SMT-EVAL-2013
105	172032.14	MathSAT5-5.2.6-SMT-EVAL-2013
111	171953.63	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
136	237594.15	SMTInterpol-2.0r8402-SMT-EVAL-2013
183	350598.34	OpenSMT-SMT-EVAL-2013
QF_UFBV 31 benchmarks , 0.79 days wall time		
0	0.32	Boolector-1.5.118-SMT-EVAL-2013
0	0.33	SONOLAR-2013-05-15-SMT-EVAL-2013
7	15962.85	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
12	23942.91	MathSAT5-5.2.6-SMT-EVAL-2013
18	28678.86	CVC4-SMT-EVAL-2013
QF_UFIDL 430 benchmarks , 1.90 days wall time		
7	11947.79	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
32	71695.10	CVC4-SMT-EVAL-2013
46	80562.09	veriT-SMT-EVAL-2013
QF_UFLIA 564 benchmarks , 2.03 days wall time		
0	100.67	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
0	355.30	MathSAT5-5.2.6-SMT-EVAL-2013
0	1090.59	CVC4-SMT-EVAL-2013
0	1469.57	SMTInterpol-2.0r8402-SMT-EVAL-2013
68	172440.25	veriT-SMT-EVAL-2013
QF_UFLRA 900 benchmarks , 0.02 days wall time		
0	40.09	MathSAT5-5.2.6-SMT-EVAL-2013
0	40.86	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
0	57.85	CVC4-SMT-EVAL-2013
0	162.95	veriT-SMT-EVAL-2013
0	1246.65	SMTInterpol-2.0r8402-SMT-EVAL-2013
QF_UFNRA 26 benchmarks , 0.12 days wall time		
0	0.25	veriT-SMT-EVAL-2013
4	10565.17	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
UFLRA 5 benchmarks , 0.11 days wall time		
0	39.29	CVC4-SMT-EVAL-2013
2	3103.69	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013
4	6300.71	veriT-SMT-EVAL-2013
UFNIA 1796 benchmarks , 1.55 days wall time		
68	133752.97	Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-SMT-EVAL-2013

Fig. 9: (Part 2) Results of a virtual competition using the 2013 solvers on all benchmarks.

For example, the logic QF_AUFBV has 14335 benchmarks. Of those, 520 are unsolved by Boolector and 543 are unsolved by SONOLAR (the winner and runner up of the virtual competition). Of these only 346 are unsolved by both, leaving 174 unsolved only by Boolector and 197 unsolved only by SONOLAR. Choosing 1433 benchmarks at random, over 1000 trials, yields more benchmarks unsolvable by Boolector about 31% of the time, more unsolved by

SONOLAR about 62% of the time, and the same about 7% of the time. Thus, the random selection of the competition benchmark set would allow SONOLAR to best Boolector about 1/3 of the time, despite Boolector being better on the overall benchmark population for this logic.

- Second, even when unsolved benchmarks are not a contributing factor, some pairs of solvers have total times that are close. Variations in time caused by slightly different choices of benchmarks might cause changes in winning order. The experiment described next explores this phenomenon further.

A second virtual competition can be run using only benchmarks that all competitors solved within the timeout period. This would not make a useful real competition because the result could be easily gamed: a solver could win by thoroughly optimizing performance on a few benchmarks and purposely not solving the remainder. However, in this evaluation, no solver has had the chance to do that; performing this evaluation allows comparing running times alone, without the additional factor of unsolved benchmarks.

Figs. 11 and 12 show the result of such a competition, for the nine 2013 solvers; it uses all benchmarks that all the solvers registered for a given logic solve. In addition we performed the same experiment of 1000 virtual competitions each using 10% of these benchmarks. Fig. 13 shows the variation in winning order across these trials. The variation is even more than that shown in Fig. 10. To obtain some insight into this variation, we tabulate, in columns 2 and 3 of Figs. 11 and 12, the mean and standard deviation⁴ of the total solution time for each of the solvers in each logic. The standard deviations are substantial compared to the differences in mean times between competing solvers; even though the distributions of solving times are not necessarily Gaussian, the data indicate a substantial probability that the order would change based solely on the random choice of benchmark subset.

3.6 Competitiveness of solvers

To assess competitiveness of the competitions and the set of solvers, we measured four quantities:

- *The ratio of the first to second place times on each benchmark in each logic.* The closer this quantity is to 1.0, the closer the race and the more the runner-up is challenging the leader. The ratio is averaged over all the benchmarks for each logic and year. The results are shown in Fig. 14: for each logic and year that had more than one competitor, the mean, and 1st, median, and 3rd quartiles of the distribution of runner-up ratios are shown.

A few logics, LRA and UFLRA, are very uncompetitive: the median case has the winner more than 10 times better than the runner-up. In a few others, such as QF_IDL, QF_NIA, QF_NRA the winner is often less than half the time of the runner-up. But in most cases, the ratio is in the 50%–100% range. That is, the races are not neck-and-neck but the winners have only a modest lead

⁴ This is the population standard deviation. To obtain the sample standard deviation scale these values by $\sqrt{0.999}$.

Logic	Benchmarks used	# Trials	Winner changed	Order changed	Avg. not solved
AUFLIA	640 / 6402	1000	0.0%	0.0%	0.213
AUFLIRA	1991 / 19917	1000	0.0%	30.9%	0.026
LRA	37 / 374	1000	4.6%	4.6%	0.261
QF_AUFBV	1433 / 14335	1000	33.6%	46.7%	0.050
QF_AUFLIA	114 / 1140	1000	24.1%	58.4%	0.042
QF_AX	55 / 551	1000	33.6%	48.0%	0.048
QF_BV	3174 / 31747	1000	2.4%	26.7%	0.043
QF_IDL	217 / 2170	1000	0.0%	0.0%	0.206
QF_LIA	588 / 5882	1000	0.3%	24.4%	0.154
QF_LRA	63 / 634	1000	13.4%	67.7%	0.038
QF_NIA	53 / 530	1000	0.0%	0.0%	0.167
QF_NRA	16 / 166	1000	0.0%	0.0%	0.183
QF_RDL	25 / 255	1000	67.8%	70.8%	0.169
QF_UF	664 / 6647	1000	8.0%	71.1%	0.016
QF_UFIDL	43 / 430	1000	0.0%	18.2%	0.066
QF_UFLIA	56 / 564	1000	1.5%	24.6%	0.024
QF_UFLRA	90 / 900	1000	63.2%	68.8%	0.000

Fig. 10: Results of virtual competitions using the 2013 solvers on random subsets of 10% of the benchmarks.

over the next finisher. Note that the particular solver that is the winner varies from benchmark to benchmark.

- *The degree to which the leader for a given benchmark changes from year to year.* Since all solvers are new each year, we placed solvers into families by the research group that produced them (e.g., the CVC family includes the CVC3 and CVC4 series of solvers). We counted it a *turnover* if the family of the winning solver changed from the previous year; we measured the fraction of benchmarks for a given logic and year that saw a turnover. The results are shown in Fig. 15. There are many cases in which there is complete turnover from one year to the next; often this is because the previous solver family no longer participated or a new solver joined the competition and dominated the results. However, overall most divisions see a more than 50% turnover from year to year. We see this as indicative of reasonably robust competition.
- *The distribution of winning solvers across the benchmarks within a logic.* A highly competitive environment would have each competing solver win a roughly equal fraction of the benchmarks; a non-competitive environment would have a single solver winning nearly all of the benchmarks. Our measure of competitiveness is a scaled entropy measure: if f_i is the fraction of benchmarks won by solver i , and there are N competing solvers, then the competitiveness measure is $2^{(-\sum_i f_i \log_2 f_i)} / N$. This quantity is p/N if the wins are equally distributed among p of the N solvers ($f_i = 1/p$ for p of the solvers and 0 for the other $N-p$). The results are shown in Figs. 16 and 17; this data only includes benchmarks that were solved by a winner. The first entropy column gives the value of $2^{(-\sum_i f_i \log_2 f_i)}$, whose value is roughly the number of solvers over which the wins are distributed. The second entropy column is $2^{(-\sum_i f_i \log_2 f_i)} / N$, which scales the first value between $1/N$ and 1.

total secs	mean for 10%	s.dev for 10%	solver
AUFLIA 4174 benchmarks , 0.22 days wall time			
134.84	13.57	1.47	Z3-4.3.2.a054b099c1d6-x64-debian-...
8572.19	881.87	725.38	veriT-SMT-EVAL-2013
10524.23	1017.33	644.67	CVC4-SMT-EVAL-2013
AUFLIRA 18699 benchmarks , 0.35 days wall time			
239.75	23.99	1.14	Z3-4.3.2.a054b099c1d6-x64-debian-...
1190.59	113.12	245.28	CVC4-SMT-EVAL-2013
29062.40	2906.81	1592.05	veriT-SMT-EVAL-2013
LRA 209 benchmarks , 0.10 days wall time			
3104.20	308.70	505.21	Z3-4.3.2.a054b099c1d6-x64-debian-...
5619.18	531.80	581.02	CVC4-SMT-EVAL-2013
QF_AUFBV 12796 benchmarks , 1.09 days wall time			
9132.25	927.34	667.35	Boolector-1.5.118-SMT-EVAL-2013
15551.14	1571.56	789.27	MathSAT5-5.2.6-SMT-EVAL-2013
18228.79	1858.63	915.22	Z3-4.3.2.a054b099c1d6-x64-debian-...
19760.57	1997.13	1010.26	SONOLAR-2013-05-15-SMT-EVAL-2013
31608.01	3146.44	1223.43	CVC4-SMT-EVAL-2013
QF_AUFLIA 1015 benchmarks , 0.07 days wall time			
33.34	3.32	1.52	veriT-SMT-EVAL-2013
93.14	9.25	11.90	Z3-4.3.2.a054b099c1d6-x64-debian-...
206.75	20.42	11.43	MathSAT5-5.2.6-SMT-EVAL-2013
5529.68	542.85	517.03	CVC4-SMT-EVAL-2013
QF_AX 491 benchmarks , 0.00 days wall time			
19.87	1.97	1.48	Z3-4.3.2.a054b099c1d6-x64-debian-...
27.64	2.76	1.28	MathSAT5-5.2.6-SMT-EVAL-2013
86.80	8.57	8.89	CVC4-SMT-EVAL-2013
QF_BV 28981 benchmarks , 5.95 days wall time			
24741.62	2466.39	574.41	Boolector-1.5.118-SMT-EVAL-2013
33090.65	3315.48	924.08	Z3-4.3.2.a054b099c1d6-x64-debian-...
58252.81	5780.94	1699.68	SONOLAR-2013-05-15-SMT-EVAL-2013
180688.61	18097.26	2555.59	CVC4-SMT-EVAL-2013
217464.30	21676.36	2996.98	MathSAT5-5.2.6-SMT-EVAL-2013
QF_IDL 1407 benchmarks , 2.45 days wall time			
10837.79	1075.06	400.21	Z3-4.3.2.a054b099c1d6-x64-debian-...
28005.20	2794.68	804.94	CVC4-SMT-EVAL-2013
173260.34	17004.68	2911.07	veriT-SMT-EVAL-2013
QF_LIA 2085 benchmarks , 1.18 days wall time			
5108.99	517.56	288.02	MathSAT5-5.2.6-SMT-EVAL-2013
9989.45	1026.38	847.97	Z3-4.3.2.a054b099c1d6-x64-debian-...
23068.05	2333.97	1019.42	CVC4-SMT-EVAL-2013
30542.46	3107.09	1214.87	veriT-SMT-EVAL-2013
33029.34	3264.05	1243.38	SMTInterpol-2.0r8402-SMT-EVAL-2013
QF_LRA 583 benchmarks , 0.60 days wall time			
1410.69	138.20	57.97	Z3-4.3.2.a054b099c1d6-x64-debian-...
2108.31	206.48	99.11	CVC4-SMT-EVAL-2013
3892.58	394.05	395.39	MathSAT5-5.2.6-SMT-EVAL-2013
21918.95	2141.70	1153.96	SMTInterpol-2.0r8402-SMT-EVAL-2013
22838.44	2236.65	1019.30	veriT-SMT-EVAL-2013
QF_NIA 357 benchmarks , 0.19 days wall time			
2474.66	233.56	284.13	MiniSMT-0.5-SMT-EVAL-2013
13683.23	1333.87	938.66	Z3-4.3.2.a054b099c1d6-x64-debian-...

Fig. 11: (Part 1) A virtual competition run only on benchmarks that all solvers solve.

total secs	mean for 10%	s.dev for 10%	solver
QF_NRA 87 benchmarks , 0.01 days wall time			
0.84			veriT-SMT-EVAL-2013
21.63			Z3-4.3.2.a054b099c1d6-x64-debian-...
548.45			MiniSMT-0.5-SMT-EVAL-2013
QF_RDL 203 benchmarks , 0.24 days wall time			
2759.19	255.65	229.84	Z3-4.3.2.a054b099c1d6-x64-debian-...
3296.74	319.74	161.95	CVC4-SMT-EVAL-2013
14350.83	1404.44	870.90	veriT-SMT-EVAL-2013
QF_UF 6332 benchmarks , 1.27 days wall time			
878.33	87.54	10.37	veriT-SMT-EVAL-2013
2740.19	266.55	285.91	Z3-4.3.2.a054b099c1d6-x64-debian-...
5374.81	525.67	428.07	MathSAT5-5.2.6-SMT-EVAL-2013
5420.78	551.93	483.45	CVC4-SMT-EVAL-2013
20575.64	2039.15	507.66	SMTInterpol-2.0r8402-SMT-EVAL-2013
74537.24	7344.50	1841.05	OpenSMT-SMT-EVAL-2013
QF_UFBV 12 benchmarks , 0.03 days wall time			
0.11			Boolector-1.5.118-SMT-EVAL-2013
0.13			SONOLAR-2013-05-15-SMT-EVAL-2013
312.28			Z3-4.3.2.a054b099c1d6-x64-debian-...
786.01			MathSAT5-5.2.6-SMT-EVAL-2013
1625.99			CVC4-SMT-EVAL-2013
QF_UFIDL 382 benchmarks , 0.26 days wall time			
385.76	40.19	32.75	Z3-4.3.2.a054b099c1d6-x64-debian-...
10350.75	1047.52	766.90	veriT-SMT-EVAL-2013
11878.04	1213.77	586.08	CVC4-SMT-EVAL-2013
QF_UFLIA 496 benchmarks , 0.83 days wall time			
66.61	7.36	14.24	Z3-4.3.2.a054b099c1d6-x64-debian-...
191.79	19.21	13.67	MathSAT5-5.2.6-SMT-EVAL-2013
628.60	64.03	58.11	CVC4-SMT-EVAL-2013
706.64	71.69	46.38	SMTInterpol-2.0r8402-SMT-EVAL-2013
70440.25	6930.04	1853.95	veriT-SMT-EVAL-2013
QF_UFLRA 900 benchmarks , 0.02 days wall time			
40.09	3.99	0.27	MathSAT5-5.2.6-SMT-EVAL-2013
40.86	4.08	0.92	Z3-4.3.2.a054b099c1d6-x64-debian-...
57.85	5.75	0.63	CVC4-SMT-EVAL-2013
162.95	16.24	12.40	veriT-SMT-EVAL-2013
1246.65	124.03	8.82	SMTInterpol-2.0r8402-SMT-EVAL-2013
QF_UFNRA 22 benchmarks , 0.05 days wall time			
0.21			veriT-SMT-EVAL-2013
4565.17			Z3-4.3.2.a054b099c1d6-x64-debian-...
UFLRA 1 benchmarks , 0.00 days wall time			
0.87			CVC4-SMT-EVAL-2013
39.07			Z3-4.3.2.a054b099c1d6-x64-debian-...
300.71			veriT-SMT-EVAL-2013

Fig. 12: (Part 2) A virtual competition run only on benchmarks that all solvers solve.

There is a trivial case of one solver, with all wins distributed equally over the $N=1$ solvers and an unscaled ‘competitiveness’ measure of 1.0; in all the combinations of logic and year, there are 18 such cases. Of the other year-logic combinations, note that in all but 11 (of the 70), despite any dominance by one solver, all competitors won at least one benchmark. The ‘competitiveness’ metric itself shows that in most cases there are approximately 2 solvers sharing the bulk of the wins, increasing to about 3 in a few cases that have

Logic	Benchmarks used	# Trials	Winner changed	Order changed
AUFLIA	417 / 4174	1000	0.0%	41.1%
AUFLIRA	1869 / 18699	1000	2.0%	3.2%
LRA	20 / 209	1000	37.4%	37.4%
QF_AUFBV	1279 / 12796	1000	31.3%	89.9%
QF_AUFLIA	101 / 1015	1000	23.3%	32.5%
QF_AX	49 / 491	1000	12.1%	35.2%
QF_BV	2898 / 28981	1000	16.3%	34.7%
QF_IDL	140 / 1407	1000	0.0%	0.0%
QF_LIA	208 / 2085	1000	35.1%	86.8%
QF_LRA	58 / 583	1000	25.9%	81.0%
QF_NIA	35 / 357	1000	12.2%	12.2%
QF_RDL	20 / 203	1000	19.0%	19.5%
QF_UF	633 / 6332	1000	0.2%	58.6%
QF_UFIDL	38 / 382	1000	0.0%	37.6%
QF_UFLIA	49 / 496	1000	7.9%	32.5%
QF_UFLRA	90 / 900	1000	60.9%	67.1%

Fig. 13: Stability of competition results when run only on benchmarks solved by all solvers.

many participants. The case of the most participants—11 solvers for QF_BV in 2012—had a distribution among about 3.5 winning solvers. Thus, although nearly all solvers contribute something, performance is dominated by a few, but not by only one, in nearly all competitive logics.

- *SOTAC*. As a final measurement in this subcategory, we measured the *state of the art contribution* (SOTAC), as proposed in [8]. This measures the uniqueness of the contribution of each solver. It does not consider the time taken to solve a benchmark, but just whether a solver solves a benchmark (within the timeout period). The contribution of a benchmark to a solver’s SOTAC is 0 if the solver does not solve the benchmark; the contribution is $1/(\text{the number of solvers that solve that benchmark})$ if it does solve it. Thus the maximum contribution is obtained when a solver is the only one to solve a benchmark. Fig. 18 shows the computation for each of the solvers for the year 2013. The first column shows the sum of the SOTAC over all benchmarks in which a solver participated; the second column is the average over all the benchmarks that that solver attempted (including ones that timed out), but not benchmarks in logics in which the solver did not participate; the third column⁵ is the average over all benchmarks for which the solver was successful (did not time out).

The CVC4 and Z3 solvers have a high total SOTAC because they contribute to a broad range of logics; MathSAT and veriT also contribute broadly. The averages in the last two columns are not significantly different from each other. MiniSMT has a high average SOTAC because it does well at just a few logics. VeriT suffers slightly on an overall average, but has a relatively higher SOTAC averaged over those benchmarks that it does solve.

⁵ The third column corresponds to Sutcliffe’s definition of SOTAC.

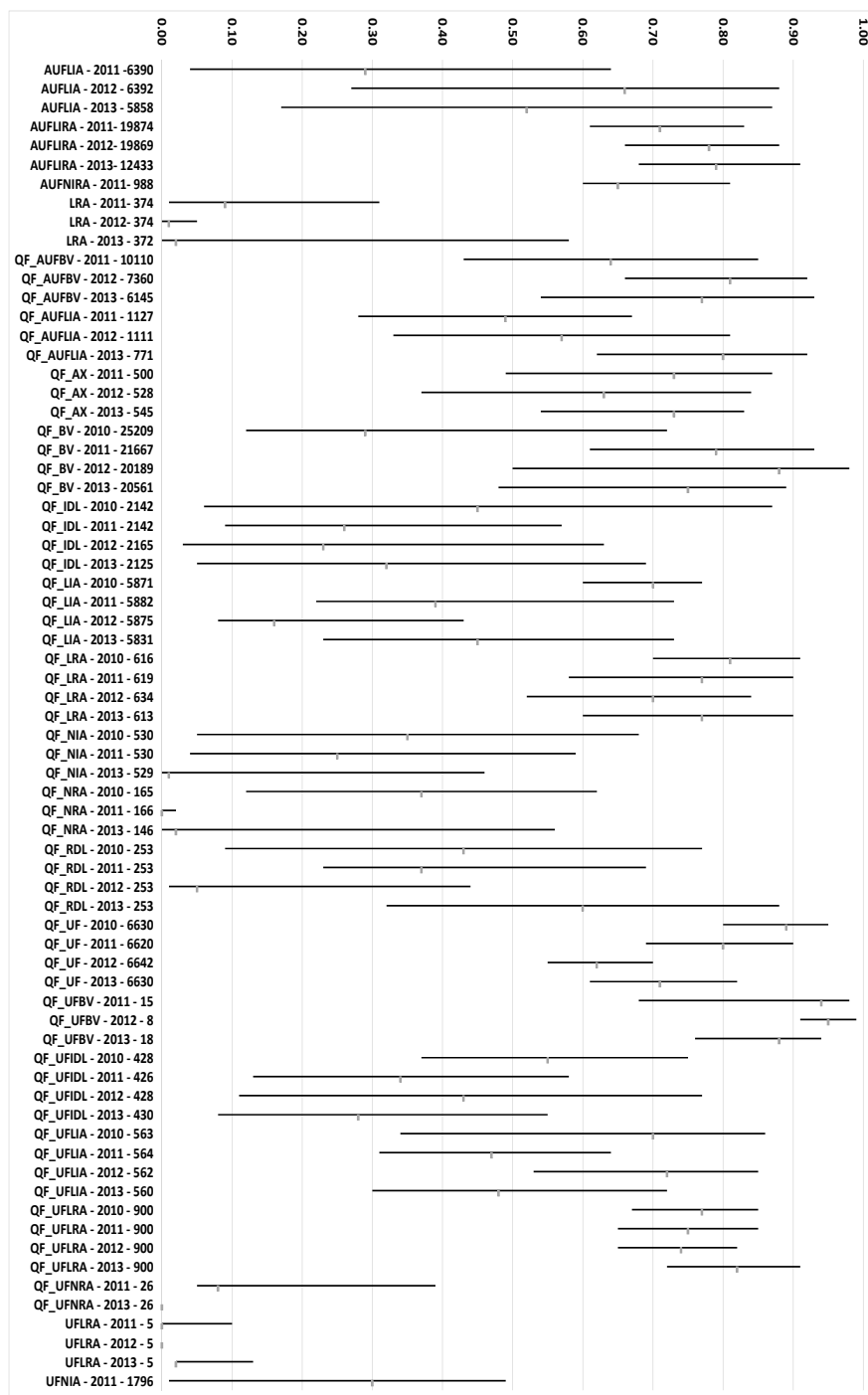


Fig. 14: Runner-up: Ratio of winning time to runner-up time, showing median, first and third quartiles.

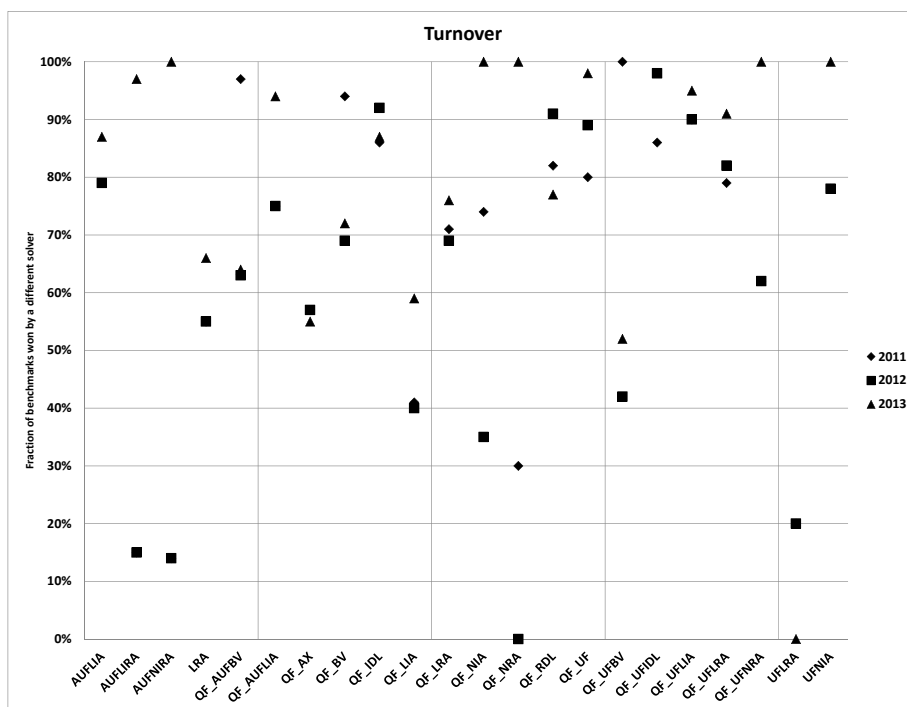


Fig. 15: Turnover: Fraction of benchmarks for the given year and logic for which the winning solver is from a different solver family than in the prior year.

3.7 Sequential vs. parallel computation

The StarExec cluster contains computation nodes with 4 cores each; solvers configured for multi-threaded computation can take advantage of the parallel processing possibilities and reduce their overall wall-clock time in solving a benchmark. Previous competitions have had demonstration divisions featuring parallel computation. For this evaluation we set a wall-clock timeout of 25 minutes and a CPU-time timeout of 100 minutes. Thus the evaluation measured the performance of solvers for parallel computation, if they were so configured. A competition focused on sequential processing would set the wall-clock and CPU-time timeouts at the same value. We did not rerun the evaluation data to assess sequential processing performance. The data showing both wall-clock and CPU time was not available when we did our primary evaluation, and some of the early CPU-time data appears inconsistent with the wall-clock times. Thus we left questions about parallel vs. sequential performance for further study:

- what fraction of solvers use parallel computation?
- what is the speedup generally achieved by these solvers and the kinds of problems represented by the benchmark set?

Logic	Year	# Solvers	# Winners	Winning dist.,	scaled	# Benchmarks
AUFLIA	2011	2	2	1.36	0.68	6056
	2012	2	2	1.61	0.81	5685
	2013	3	3	1.69	0.56	5771
AUFLIRA	2011	2	2	1.34	0.67	19791
	2012	2	2	1.42	0.71	19753
	2013	3	3	1.73	0.58	19845
AUFNIRA	2011	2	2	1.32	0.66	987
LRA	2011	2	2	1.58	0.79	322
	2012	2	2	1.02	0.51	358
	2013	2	2	1.52	0.76	344
QF_AUFBV	2011	5	5	2.16	0.43	14027
	2012	6	6	2.44	0.41	14124
	2013	5	5	2.55	0.51	14128
QF_AUFLIA	2011	3	3	1.60	0.53	1126
	2012	3	3	2.01	0.67	1118
	2013	4	4	1.84	0.46	1131
QF_AX	2011	4	4	2.27	0.57	547
	2012	2	2	1.56	0.78	541
	2013	3	3	1.87	0.62	548
QF_BV	2010	3	3	1.71	0.57	31094
	2011	8	8	3.02	0.38	31320
	2012	11	11	3.46	0.31	31369
	2013	5	5	2.36	0.47	31230
QF_IDL	2010	3	3	1.87	0.62	1756
	2011	4	4	1.36	0.34	1954
	2012	2	2	1.15	0.57	1830
	2013	3	3	1.41	0.47	1956
QF_LIA	2010	3	3	1.45	0.48	5833
	2011	4	4	1.83	0.46	5829
	2012	4	4	1.44	0.36	5789
	2013	5	5	2.38	0.48	5859
QF_LRA	2010	5	5	2.10	0.42	618
	2011	6	6	2.42	0.40	620
	2012	4	4	1.63	0.41	619
	2013	5	5	2.42	0.48	627

Fig. 16: (Part 1) Winner distribution: The degree to which winning is distributed among solvers vs. dominated by one solver.

- what is the observed overhead, that is, for purely sequential performance by how much does the wall-clock time exceed the CPU-time?
- are competition results significantly different when judged by CPU time vs. wall-clock time?

We expect that the 2014 competition will generate a data set better able to answer these questions.

With respect to this study, note that we used solvers as they were historically configured. They may have been configured solely for sequential performance, and the authors do not know in every case what the configurations of the older solvers are. Modern configurations of the same solvers might make use of multiple cores. The configuration of StarExec used for the evaluation (4 cores, 25 minute time-out) might be less than ideal for some historical solvers, but it is typical of what

Logic	Year	# Solvers	# Winners	Winning dist.,	scaled	# Benchmarks
QF_NIA	2010	3	3	1.64	0.55	530
	2011	3	3	1.78	0.59	530
	2013	2	2	1.58	0.79	526
QF_NRA	2010	2	2	1.52	0.76	166
	2011	2	1	1.00	0.50	166
	2013	3	3	1.17	0.39	166
QF_RDL	2010	3	3	1.61	0.54	220
	2011	4	3	1.32	0.33	222
	2012	2	2	1.04	0.52	213
	2013	3	3	1.80	0.60	218
QF_UF	2010	5	5	2.15	0.43	6643
	2011	7	7	2.29	0.33	6647
	2012	4	4	1.57	0.39	6643
	2013	6	6	1.71	0.28	6647
QF_UFBV	2011	4	2	1.59	0.40	31
	2012	5	3	2.03	0.41	31
	2013	5	3	1.75	0.35	31
QF_UFIDL	2010	3	3	1.81	0.60	422
	2011	4	4	1.43	0.36	423
	2012	2	2	1.41	0.70	404
	2013	3	3	1.32	0.44	423
QF_UFLIA	2010	3	3	2.08	0.69	564
	2011	4	4	1.41	0.35	564
	2012	4	4	1.88	0.47	564
	2013	5	4	1.70	0.34	564
QF_UFLRA	2010	3	3	1.38	0.46	900
	2011	4	2	1.43	0.36	900
	2012	4	4	1.29	0.32	900
	2013	5	4	1.95	0.39	900
QF_UFNRA	2011	2	2	1.59	0.79	26
	2013	2	1	1.00	0.50	26
UFLRA	2011	2	2	1.41	0.71	5
	2012	2	1	1.00	0.50	5
	2013	3	1	1.00	0.33	5
UFNIA	2011	2	2	1.41	0.71	1732

Fig. 17: (Part 2) Winner distribution: The degree to which winning is distributed among solvers vs. dominated by one solver.

Solver	Total SOTAC	Mean over all attempted	Mean over all solved
Z3-4.3.2.a054b099c1d6-x64-debian-6.0.6-...	25664.33	0.27	0.28
CVC4-SMT-EVAL-2013	21562.27	0.23	0.25
MathSAT5-5.2.6-SMT-EVAL-2013	12285.92	0.20	0.21
veriT-SMT-EVAL-2013	11446.57	0.25	0.30
Boolector-1.5.118-SMT-EVAL-2013	9352.38	0.20	0.21
SONOLAR-2013-05-15-SMT-EVAL-2013	9031.47	0.20	0.21
SMTInterpol-2.0r8402-SMT-EVAL-2013	2829.63	0.19	0.20
OpenSMT-SMT-EVAL-2013	1083.43	0.16	0.17
MiniSMT-0.5-SMT-EVAL-2013	397.00	0.57	0.60

Fig. 18: State of the art contribution from each of the 2013 solvers, sorted by total SOTAC.

a user might use at the present time. Thus, to repeat a point made before in Section 2.2, the results of the virtual competitions should not be used to compare specific solvers. Rather, they offer a comparison of different modes of organizing and running the competition.

3.8 Benchmarks

The number of benchmarks has successfully grown since SMT-LIB was established; at the time of this evaluation, there were more than 100 000 in all of the logics combined, including both incremental and non-incremental collections. However, the logics differ significantly in the number of available benchmarks and in their overall difficulty.

SMT-LIB distinguishes four loosely-defined kinds of benchmarks:

- *check* benchmarks are simple tasks that are designed to ensure that a solver has the basic functionality required for a division;
- *industrial* benchmarks are generated from some application; ideally these are substantial examples showing real-world variation, but they may be from toy applications running on a toy examples;
- *crafted* benchmarks are hand-crafted to exercise a particular functionality or technical challenge;
- *random* benchmarks are randomly generated from some distribution.

Fig. 19 shows the number of benchmarks for each logic and the distribution over the categories just listed. The numbers vary considerably. Some logics, such as UFLRA, have just a very few benchmarks, while others have tens of thousands. Note that in nearly all logics, the bulk of the benchmarks are considered industrial.

The same table also shows the distribution of benchmarks among those known to be satisfiable, unsatisfiable, and with an unknown status. Here there is considerable variation. Some logics have nearly no satisfiable benchmarks. Some have a large proportion (at least as of 2013) of benchmarks of unknown status. This is particularly the case for the popular and important QF_BV logic. Part of the effort of benchmark curation for the 2014 competition is to more carefully assign benchmarks to logics and to determine status that is currently unknown.

Some benchmarks are *incremental*; that is, they contain more than one **check-sat** command in a command script. This is relevant to interactive applications, but is not the main focus of the competition. In 2012, there were demonstration divisions on generating unsat cores and proof generation. These did not require special benchmarks, but do require different evaluation. The evaluation reported here did not consider the incremental benchmarks.

Fig. 20 shows the fraction of benchmarks (for each logic and year) that are solved within the 25 minute timeout period. For each combination of solver and year we report the fraction of benchmarks that are completed by all solvers and the fraction that are completed by at least one solver. In 2013, in about half of the logics, all benchmarks are completed by at least one solver within the timeout period (though not necessarily the same solver); in all but one logic (QF_RDL),

Logic	Total	industrial	crafted	random	check	sat	unsat	unknown
AUFLIA	6402	6375	26	0	128	9	4149	2244
AUFLIRA	19917	19717	200	0	0	100	19742	75
AUFNIRA	989	989	0	0	0	0	943	46
LRA	374	144	0	230	0	0	0	374
QF_AUFBV	14335	14008	325	0	16	4795	3566	5974
QF_AUFLIA	1140	114	1020	0	6	548	592	0
QF_AX	551	0	551	0	0	272	279	0
QF_BV	31748	29887	1160	3	10	2259	1378	28111
QF_IDL	2170	628	980	63	2	942	691	537
QF_LIA	5882	3832	737	1075	1118	2227	2584	1071
QF_LRA	634	632	0	0	2	226	317	91
QF_NIA	530	177	167	0	0	242	63	225
QF_NRA	166	151	15	0	0	56	4	106
QF_RDL	255	147	106	0	2	57	113	85
QF_UF	6647	0	6647	0	0	2549	4094	4
QF_UFBV	31	31	0	0	0	0	31	0
QF_UFIDL	430	396	32	0	2	110	318	2
QF_UFLIA	564	331	231	0	2	396	168	0
QF_UFLRA	900	0	0	900	0	409	407	84
QF_UFNRA	26	26	0	0	0	15	0	11
UFLRA	5	0	5	0	0	5	0	0
UFNIA	1796	1796	0	0	0	0	1660	136
Total	95492	79381	12202	2271	1288	15217	41099	39176

Fig. 19: The numbers of benchmarks by logic, category, and result.

at least 95% of the benchmarks are completed by at least one solver. The values for the fraction of benchmarks completed by all solvers are not as high, since some of the solvers may be initial experimental versions and not tuned for competition. Even so, in more than half of the logics, at least 85% of benchmarks were completed by all solvers.

Fig. 21 is another view of benchmark difficulty. Here, for each logic, the distributions of winning times among the 2013 solvers are shown. In particular, selected percentiles of each distribution are tabulated (the value for the n th percentile is the number of seconds for which that fraction of the benchmarks is completed by the winning solver for that benchmark). In all but three logics more than 80% of the benchmarks for that logic take less than just a few seconds, if not less than a second. Only four of the logics have more than 5% of their benchmarks that take more than the timeout period.

These results indicate that there is substantial room for more difficult benchmarks in almost all of the logics. On the other hand, if the benchmarks are indeed typical of application problems, the results indicate that today's solvers are very capable.

An additional question that would be interesting to address in a benchmark evaluation is the degree to which benchmarks discriminate among solvers. A small set of 'discriminating' benchmarks would be more useful than a large set of undifferentiated ones. One could, for example, measure a quantity akin to the SOTAC measure of Section 3.6: benchmarks that are easy for all solvers would be discounted, ones that are hard for all would be challenges, and ones that show a wide range of performance might be discriminating benchmarks.

Logic	#	2010 all	2011 all	2012 all	2013 all	2010 any	2011 any	2012 any	2013 any
AUFLIA	6402	0.80	0.75	0.73	0.65	0.80	0.95	0.89	0.90
AUFLIRA	19917	0.98	0.97	0.95	0.94	0.98	0.99	0.99	1.00
AUFNIRA	989	0.98	0.98	0.99	0.99	0.98	1.00	0.99	0.99
LRA	374	0.78	0.57	0.73	0.56	0.78	0.86	0.96	0.92
QF_AUFBV	14335	0.85	0.79	0.80	0.89	0.85	0.98	0.99	0.99
QF_AUFLIA	1140	0.93	0.86	0.85	0.89	0.93	0.99	0.98	0.99
QF_AX	551	0.94	0.86	0.90	0.89	0.94	0.99	0.98	0.99
QF_BV	31747	0.87	0.84	0.82	0.91	0.98	0.99	0.99	0.98
QF_IDL	2170	0.44	0.47	0.50	0.65	0.81	0.90	0.84	0.90
QF_LIA	5882	0.80	0.24	0.33	0.35	0.99	0.99	0.98	1.00
QF_LRA	634	0.74	0.73	0.73	0.92	0.97	0.98	0.98	0.99
QF_NIA	530	0.78	0.56	0.98	0.67	1.00	1.00	0.98	0.99
QF_NRA	166	0.93	0.46	1.00	0.52	1.00	1.00	1.00	1.00
QF_RDL	255	0.52	0.51	0.52	0.80	0.86	0.87	0.84	0.85
QF_UF	6647	0.96	0.95	0.96	0.95	1.00	1.00	1.00	1.00
QF_UFBV	31	1.00	0.26	0.42	0.39	1.00	1.00	1.00	1.00
QF_UFIDL	430	0.73	0.73	0.72	0.89	0.98	0.98	0.94	0.98
QF_UFLIA	564	0.91	0.91	0.91	0.88	1.00	1.00	1.00	1.00
QF_UFLRA	900	0.89	0.61	0.61	1.00	1.00	1.00	1.00	1.00
QF_UFNRA	26	1.00	0.96	1.00	0.85	1.00	1.00	1.00	1.00
UFLRA	5	1.00	0.60	0.40	0.20	1.00	1.00	1.00	1.00
UFNIA	1796	0.63	0.74	0.74	0.96	0.63	0.96	0.74	0.96

Fig. 20: The fraction of benchmarks completed by the solvers for the given year and logic.

For example, consider a competition with just two solvers. There are then four categories of benchmarks corresponding to the subsets of solvers that solve the benchmarks. The key point is whether benchmarks that are grouped together would nearly always be grouped together in other competitions with other sets of solvers or other competition rules. To fully elucidate this question will require analysis of the characteristics of the benchmarks combined with experimental assessment of their similarity in a wide variety of competition designs. We determined that the principal data collection of SMT-EVAL was not adequate to answer this question and that it deserved a full analysis in conjunction with a broad evaluation of benchmarks.

3.9 Application needs

As part of its evaluation, the SMT-EVAL team solicited input on applications that use SMT-LIB. The response was not broad enough to be representative. In addition, the authors have encountered, by happenstance, enough users of SMT-LIB who are not active in the user community to indicate that there is likely a wide variety of uses that are not well-organized or well-represented in benchmarks on specific logics. A few application domains are fairly well-known, including software verification and synthesis, scheduling, planning and routing optimization, and invariant inference. Software verification applications in particular will benefit from broader support for combinations of theories and better heuristics for quantification.

Logic	#	Median	80th percentile	90th percentile	95th percentile
AUFLIA	6402	0.02	0.07	7.26	1600.00
AUFLIRA	19917	0.01	0.01	0.01	0.02
AUFNIRA	989	0.01	0.01	0.02	0.02
LRA	374	0.02	0.07	2.81	1600.00
QF_AUFBV	14335	0.01	0.02	0.05	0.16
QF_AUFLIA	1140	0.01	0.01	0.02	0.04
QF_AX	551	0.02	0.03	0.05	0.07
QF_BV	31747	0.01	0.15	0.58	1.06
QF_IDL	2170	0.74	28.61	1355.54	1600.00
QF_LIA	5882	0.30	3.91	11.97	19.02
QF_LRA	634	0.09	1.12	8.89	25.38
QF_NIA	530	0.02	0.15	0.69	1.61
QF_NRA	166	0.01	0.01	0.01	0.01
QF_RDL	255	1.76	41.93	1600.00	1600.00
QF_UF	6647	0.03	0.12	0.28	0.48
QF_UFBV	31	0.01	0.01	0.01	0.01
QF_UFIDL	430	0.13	0.67	2.28	15.11
QF_UFLIA	564	0.02	0.03	0.05	0.10
QF_UFLRA	900	0.03	0.04	0.05	0.06
QF_UFNRA	26	0.01	0.01	0.01	0.01
UFLRA	5	3.30	28.19	28.19	28.19
UFNIA	1796	0.22	2.40	59.47	383.20

Fig. 21: The distribution of winning benchmark times by logic (for 2013 solvers).

In general, a better understanding of the variety of application needs is needed to target future development of SMT solvers and the SMT benchmark library.

4 Conclusions and recommendations

4.1 Observations from the evaluation data

The analyses described point to three principal conclusions and a number of observations.

- First, unsurprisingly, there is still a need for more and better benchmarks, despite the successful growth and current large quantity ($> 100\,000$) of benchmarks in SMT-LIB. Some logics have few benchmarks. In most logics, only a small fraction of benchmarks are significantly challenging (measured by time required to solve them). Effort toward determining status of benchmarks whose status is currently unknown would also be worthwhile.
- Related to the above, a better sense of the application areas of SMT is needed. Such an analysis will drive solver research in application-oriented directions, provide focus on application-oriented logics, and guide application-relevant benchmark acquisition.
- Finally, we discovered that using a random subset of benchmarks as the basis for the annual competition, together with inherent non-determinism in running solvers, significantly lessens the ability of a competition to determine ‘best’ solvers at a given point in time. Simply rerunning a competition is quite likely

to result in a different ordering of results. The best mitigation is to run as large a benchmark set as possible in a competition (and to work toward application-relevant benchmark sets).

Other observations are these:

- *Participation.* The number of participants is relatively stable (9 to 13) with an average turnover of 35% (low of 16%, high of 50%) each year. There is also a core of continuing participants (about 1/3 of the total).
- *Progress in solver performance.* Solver performance increased significantly from 2010 to 2013. Most of that improvement occurred from 2010 to 2011; we did not determine a reason for that improvement. There is improvement across nearly all logics, though some logics are lagging.
- *Repeatability.* The repeatability of benchmarks is reduced by solver nondeterminism. Most importantly, however, competition repeatability is compromised by significant differences in performance of solvers on particular benchmarks, such that different random subsets of benchmarks will produce different assessments of solvers. The relevance of competitions is also threatened by the relevance of the total benchmark set.
- *Competitiveness of solvers.* Within a logic nearly all solvers contribute something (e.g., a solution that no other solver produces). However, generally the winning times across benchmarks within a logic are obtained by 2-3 solvers that dominate that division. Comparing first-place times to second-place times shows that (a) in many logics the runner-up is quite close to the winner, but (b) in some logics, presumably with state-of-the-art tuning and advanced algorithms, the winner far outstrips the other competitors. It is also relevant that most benchmarks change winner from one year to the next. Overall, we judge that there is a moderate degree of competitiveness among solvers - a few tend to dominate in any logic, but other solvers do also make their contribution.
- *Difficulty of benchmarks.* A relatively low fraction of benchmarks require more than the timeout period to solve. Most benchmarks are solved by all solvers within the timeout period.

4.2 Recommendations for competitions and for SMT-LIB

The experience of past competitions and this evaluation point to a number of ideas for future competitions. Most past policies have worked well and should be continued: openness, transparency, reproducibility, public submission of solvers. Other aspects could use improvement. Some of these have already begun for 2014.

- To reduce potential variation caused by choice of benchmarks in a competition, a competition should include as many benchmarks as possible.
- Invigorate the collection of benchmarks and promote a round of curating the existing benchmarks. This might include discontinuing or combining some logics.

- Encourage the participation of new entrants through tool support and recognition. Some ideas are recognizing the best new entrant and promoting reference infrastructure for elements like parsing input files or standard reporting mechanisms.
- Encourage broad participation by measuring performance in all logics, even when some logics are deemed more relevant than others.
- Encourage broadening measures of performance beyond solution of single satisfiability problems, by including, for example, determining unsat cores, measuring performance on incremental benchmarks, computing interpolants, producing proofs, producing satisfying assignments, or encouraging solvers that use multi-processing implementations.
- Investigate the naming convention and inclusion relationships for logics, with the goal of simplifying the relationships and with an eye toward adding additional theories and a significant expansion of the number of logics. This may not mean changing existing names, but would at least set a policy for how new theories and logics are to be named. At minimum this would clarify the process of naming new theories and logics.
- There is a relative paucity of difficult benchmarks (those that are not solved within the timeout period). Perhaps this is because of progress in solver performance and in hardware speed. In either case, additional difficult benchmarks are needed.
- The scoring criterion should be revisited. The current scoring emphasizes number solved over time to solve. First, time performance is important to applications, perhaps even more than solving the last 10% of hard problems, so some balance between number solved and time to solve might be appropriate. Second, some solvers issue ‘unknown’ indicating a possible satisfying solution is available, but it is not known to be sound (perhaps because of the presence of quantifiers). Such a solution may be available well before a timeout; however, such a result currently counts the same as a timeout.

4.3 Future work

The most important aspect of future work is to increase and improve benchmark quality, as described above. Thus we would encourage

- a broad survey of SMT application domains and their needs in solver support, both logics and functionality beyond determining satisfiability. A corresponding review of the benchmarks in SMT-LIB would determine the degree to which benchmarks are representative of the application domains and would provide guidelines for collecting new benchmarks.

There are also a number of questions that we posed but did not have the time or data to answer.

- Are there interesting differences in performance between satisfiable and unsatisfiable benchmarks?

- Do solvers support all of SMT-LIB v2? Should the community encourage expansion to new capabilities (e.g., unsat cores, incremental, interpolants, model generation, proof generation)? Should the community define a core language that is used for the competition, with other aspects being optional?
- Which logics should be retired or deemphasized, based on solver progress or application need?
- Competitions primarily measure ability to solve benchmarks, with the time to do so a secondary criterion. However, in some applications it is important to discharge easy problems quickly. We did not assess this characteristic of the available solvers, but it would be worth doing so.
- It would also be worth extending our study with an evaluation of incremental benchmarks and means of measuring performance in an interactive environment.
- Measuring the contribution of individual benchmarks.

As this report is being finalized, the 2014 SMT Competition is underway. Some of the proposals and hopes expressed in this report have been realized: 2014 is seeing a record number of participants, including new solver entries; there is a record number of benchmarks being used; and with only minor, recoverable problems, the StarExec infrastructure is performing a massive amount of computation over a week or so of competition time.

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