

ARCH-COMP20 Repeatability Evaluation Report

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Abstract

This report presents the results of the repeatability evaluation for the 5th International Competition on Verifying Continuous and Hybrid Systems (ARCH-COMP'21). The competition took place as part of the workshop Appplied Verification for Continuous and Hybrid Systems (ARCH) in 2021, affiliated with the 7th IFAC Conference on Analysis and Design of Hybrid Systems (ADHS'21). In its fifth edition, seventeen tools submitted artifacts through a Git repository for the repeatability evaluation, applied to solve benchmark problems for seven competition categories. The majority of participants adhered to the requirements for this year's repeatability evaluation, namely to submit scripts to automatically install and execute tools in containerized virtual environments (specifically Dockerfiles to execute within Docker), and several categories used performance evaluation information from a common execution platform. The repeatability results represent a snapshot of the current landscape of tools and the types of benchmarks for which they are particularly suited and for which others may repeat their analyses. Due to the diversity of problems in verification of continuous and hybrid systems, as well as basing on standard practice in repeatability evaluations, we evaluate the tools with pass and/or failing being repeatable.

1 Introduction

This report summarizes the *repeatability evaluation for verification of continuous and hybrid systems* for the ARCH-COMP friendly competition held in conjunction with the ARCH workshop, and aims to provide an overview of the usability and reproducibility of results for the participating verification tools. The verification community publishes papers that emphasize computational contributions, but subsequent re-creation of these computational elements is often challenging because details of the implementation are unavoidably absent in the papers. To address this challenge, some authors post code and data to their websites, but there is often limited formal incentive to do so, and typically there is no easy way to determine whether others can actually use or extend the results. Owing to such factors, computational results often become non-reproducible, sometimes even by the researchers who originally produced them.

Over about the past decade and increasingly in the past few years, the community has instituted artifact evaluations and repeatability evaluations in various phases of review processes to address these issues. The goal of the repeatability evaluation for ARCH-COMP is to improve the reproducibility of computational results for the tools competing on the selected benchmarks evaluated in the competition and to provide further confidence in the results.

This report presents a summary of the repeatability evaluation (RE) results obtained in the 2021 friendly competition of the ARCH workshop¹. The results obtained in the competition have been verified by an independent repeatability evaluation conducted by the author of this report. To establish further confidence in the results, the artifacts, code, documentation, benchmarks, etc. with which the repeatability results have been obtained are publicly available on the ARCH website (<https://cps-vo.org/group/ARCH>) and a Git version control repository (<https://gitlab.com/goranf/ARCH-COMP>).

The repeatability evaluation of the competition featured seven categories and 17 software tools, where several tools participated in multiple categories, but have been counted distinctly for their participation in each category. The categories of problems that tools participated in the repeatability evaluation are:

- AFF: affine and piecewise affine dynamics (4 tools),
- AINNCS: artificial intelligence and neural network control systems (3 tools),
- HSTP: hybrid systems theorem proving (2 tools),
- NLN: nonlinear dynamics (5 tools), and
- SM: stochastic models (3 tools).

The tools evaluated, broken into their competition categories are:

- AFF
 - CORA [1],
 - SpaceEx [10],
 - HyDRA [18], and
 - JuliaReach [7].
- AINNCS
 - NNV [25, 24, 20, 21],
 - JuliaReach [7], and
 - Verisig [13].
- HSTP
 - HHL Prover [23] and
 - KeYmaera X [17, 11].
- NLN
 - Ariadne [4, 5],
 - CORA [1],
 - Dynibex [9],
 - Kaa [14], and
 - JuliaReach [6].
- SM

¹Workshop on Appplied Verification for Continuous and Hybrid Systems (ARCH), cps-vo.org/group/ARCH

- AMYTISS [15],
- Mascot-SDS [12, 16], and
- SReachTools [22].

Some tools may have participated in the competition, but did not participate in the repeatability evaluation, so only those that participated in repeatability are listed. In future iterations, we encourage all participants of the competition to complete the repeatability evaluation to make it easier for others in the research community to build on these results, and are considering requiring repeatability participation in the future.

2 Repeatability Evaluation Plan, Execution, and Results

The repeatability evaluation was conducted primarily before and partially following the presentations of the competition results at the ARCH’21 workshop. The basic mechanism followed in the repeatability evaluation was similar to that done in related conferences. Three basic criteria are generally evaluated: coverage, instructions, and quality, each of which may be rated on a scale of one through five, where one indicates a missing component or significantly below acceptability, and five indicates the criteria significantly exceeds expectations. Coverage measures the repeatability packages’ ability to regenerate the images, tables, and log files presented in the competition. Instructions measures the packages’ ability to describe to another researcher how to reproduce the results, including installation of the tool and how to execute it. Quality measures the packages’ level of documentation and trustworthiness of results with respect to the quality of the software tool and the results it produces. This report does not describe the ratings of these review criteria for each tool evaluated, only the aggregate result of whether the submission was repeatable or not.

The participants were sent instructions to provide their tool setup instructions and tool execution commands for the benchmarks evaluated in their respective categories, which were collected on a Git repository (<https://gitlab.com/goranf/ARCH-COMP>) by the competitors issuing commits and subsequent pull/merge requests that were reviewed and approved by the author of this report. The repeatability evaluation was performed on the competition benchmarks, the selection of which has been conducted within the forum of the ARCH website (cps-vo.org/group/ARCH), which is visible for registered users and registration is open for anyone.

For all the tools listed above, which are those participating in the repeatability evaluation, all were evaluated to have passed the repeatability evaluation with their benchmark analysis results deemed repeatable. The repeatability evaluation was conducted by the author, and took approximately two weeks to complete. As in the last iteration repeatability evaluation at ARCH-COMP20, the usage of Docker significantly simplified the repeatability and we strongly encourage using this type of mechanism for repeatability evaluations, relative to earlier efforts where the evaluation was conducted primarily on a VMWare virtual machine by installing and executing all the tools. The majority of the tool authors used Docker by providing Dockerfiles, and also provided a script to execute their tool with appropriate parameters for all the benchmarks. All tools that provided Dockerfiles were able to be installed by setting up the Docker containers, then executed by the author with their provided instructions, but the author interacted with some tool developers for additional instruction for installing, executing, and/or plotting their results, in some cases interacting through the version control repository. The host machine ($M_{\text{Repeatability_Host}}$) used for executing the tools and benchmarks was an Amazon EC2 g4dn.4xlarge instance.

Category	Tool	Dockerfile?	Execution Scripts?	Performance Evaluation?
AFF	C2E2	Yes	Yes	Yes
	CORA	Yes	Yes	Yes
	Hydra	Yes	Yes	Yes
	Hylaa	Yes	Yes	Yes
	JuliaReach	Yes	Yes	Yes
	SpaceEx	Yes	Yes	Yes
	XSpeed	Yes	Yes	Yes
AINNCS	NNV	Yes	Yes	No
	JuliaReach	Yes	Yes	No
	Verisig	No	Yes	No
HSTP	HHL	Yes	Yes	No
	KeYmaera X	Yes	Yes	No
NLN	Ariadne	Yes	Yes	Yes
	CORA	No	Yes	Yes
	Dynibex	Yes	Yes	Yes
	Kaa	Yes	Yes	Yes
	JuliaReach	Yes	Yes	Yes
SM	AMYTESS	Yes	Yes	No
	FAUST ²	No	Yes	No
	Mascot-SDS	No	Yes	No
	SReachTools	No	Yes	No
	StocHy	No	Yes	No

Table 1: Summary of repeatability artifacts for each category and tool that participated in the evaluation.

As begun last year, several categories provided batch execution scripts that would execute all tools on all benchmarks in a given category, with a standardization process conducted on the CPS-VO forums for the output format to generate performance comparison tables in the individual category reports. This process in particular had a few difficulties as it only had been tested in most cases when attempting the repeatability evaluation, but most issues were resolved, and several categories (AFF, NLN) presented performance evaluation results generated for the repeatability evaluation. Overall, the tool developers provided sufficient information to install, execute, and repeat the results they obtained in the competition, although there were some issues with installation, such as missing dependencies or incompatible library versions.

3 Conclusion and Outlook

This report summarizes the repeatability evaluation for the fifth competition for the formal verification of continuous and hybrid systems (ARCH-COMP’21), conducted as part of the ARCH’21 workshop at the 7th IFAC Conference on Analysis and Design of Hybrid Systems (ADHS’21). Detailed reports for the categories can be found in the proceedings (<https://cps-vo.org/group/ARCH/proceedings>) and on the ARCH website (<http://cps-vo.org/group/ARCH>). All documentation, benchmarks, and execution scripts for the repeatability evaluation are also archived on the ARCH website, and authors contributed their repeatability evaluations to the Git repository: <https://gitlab.com/goranf/ARCH-COMP>.

As in previous iterations of the competition and corresponding repeatability evaluation, several aspects to improve the process were identified. In particular, there are still needs for (1) greater standardization of input formats, (2) standardization of output formats and results, and (3) increased execution in a common computational platform so that results, particularly performance metrics, are more meaningful. Of these challenges, this iteration of the repeatability evaluation improved upon the standardization of output formats and results, and execution on a common computational platform.

For future competitions and repeatability evaluations, several factors may still be improved by the community in future competitions. While the relatively common input format of SpaceEx in part via HyST [2] provides some means for standardizing problem specifications, there is still a greater need for utilizing a common language for specifying models and specifications. Future participants may make further use of the HyST design studio on the CPS-VO to address this issue (<https://cps-vo.org/group/hyst>). For some categories though, there are more fundamental issues with input formats. Particularly, for the stochastic models category, there are currently no standardized formats, so effort is highly recommended to address this standardization, although this area is even more challenging than non-stochastic hybrid systems, as there are many ways to model sources of uncertainty (such as through stochastic transitions a la Markov chain transitions, continuous uncertainty with stochastic differential equations, etc.). Similarly, for the AINNCS category, standardization of formats for representing both plants (e.g. as SpaceEx) and machine learning components (e.g., neural networks) should be pursued, and for the neural networks, recent efforts such as the Open Neural Network Exchange (ONNX) format or the more recent formalization of neural network semantics and specifications such as VNN-LIB (<http://www.vnnlib.org/>) should be leveraged, and taking advantage of lessons learned in the Verification of Neural Networks Competition (VNN-COMP, <https://sites.google.com/view/vnn2021>) [3]. As has been the case in past iterations, providing the ability to specify comparable parameters across different tools, as well as the particular problem domain/category (verification vs. falsification, etc.), remains a major challenge.

Second, a greater challenge still remains compared to standardizing inputs, is determining more quantitative means to compare the output results of the tools, although some libraries for common representations of reachable sets are starting to become available that may aid this process in the future, such as HyPro [18]. Figures of reachable sets and yes/no/maybe verified results for a given specification are means to make comparisons currently, but developing and standardizing a common output format may provide increased benefits and improve the ability to make quantitative comparisons between methods and tools.

Third, while this year continued for the second time performance comparisons in several categories, this remains a significant challenge for the repeatability evaluation to also repeat the performance results. Beyond these suggested improvements, there are still numerous aspects to improve, but in part through this competition and evaluation, our efforts may serve to enhance the reproducibility of computational results and increase the scientific rigor in the community.

4 Acknowledgments

The material presented in this report is based upon work supported by the National Science Foundation (NSF) under grant number FMITF 1918450 and the Defense Advanced Research Projects Agency (DARPA) Assured Autonomy program through contract number FA8750-18-C-0089. The U.S. Government is authorized to reproduce and distribute reprints for Government purposes notwithstanding any copyright notation thereon. The views and conclusions contained herein are those of the authors and should not be interpreted as necessarily representing the official policies or endorsements, either expressed or implied, of DARPA or NSF.

A Specifications of Used Machines

A.1 $M_{\text{Repeatability_Host}}$

- Amazon EC2 Instance Type: g4dn.4xlarge
- Processor: Intel Xeon Scalable (2nd Generation Cascade Lake), 16 vCPUs (AWS/EC2 Custom), 2.5 GHz base, roughly Xeon Gold 5200 Series with 24 physical cores
- Memory: 64GB
- Average CPU Mark on www.cpubenchmark.net: 25740 (full), 2396 (single thread) (for comparable Xeon Gold 5200 series)
- Host Operating System: Ubuntu

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