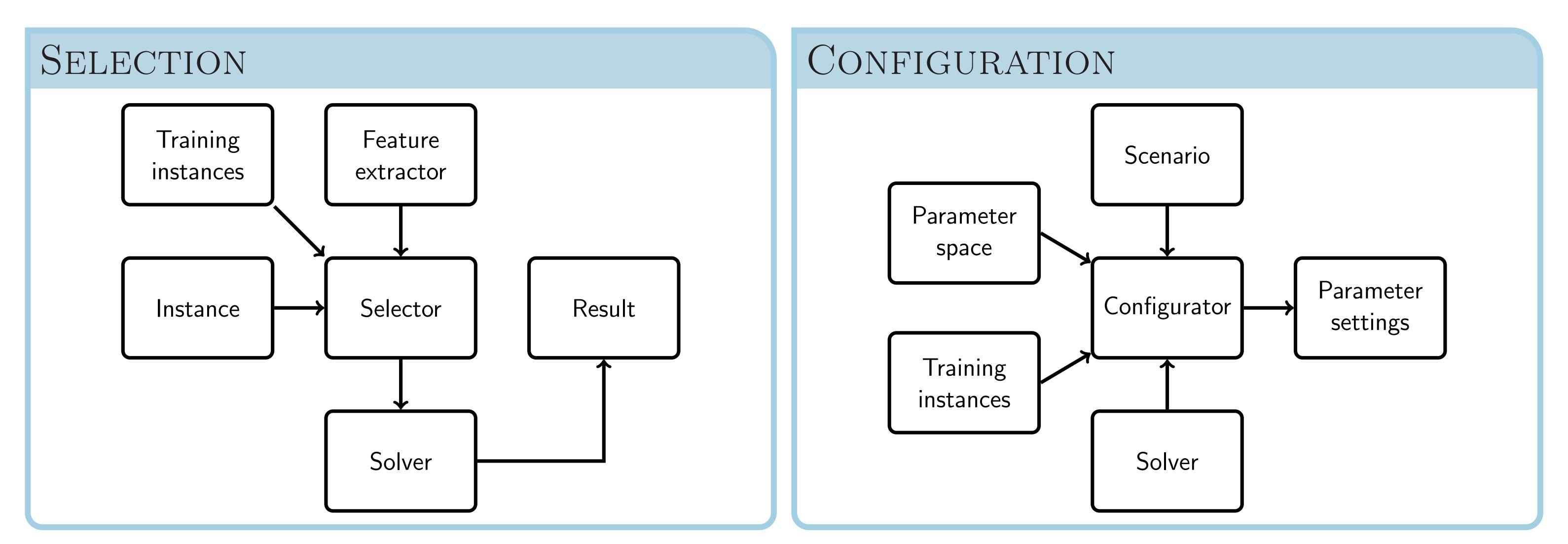
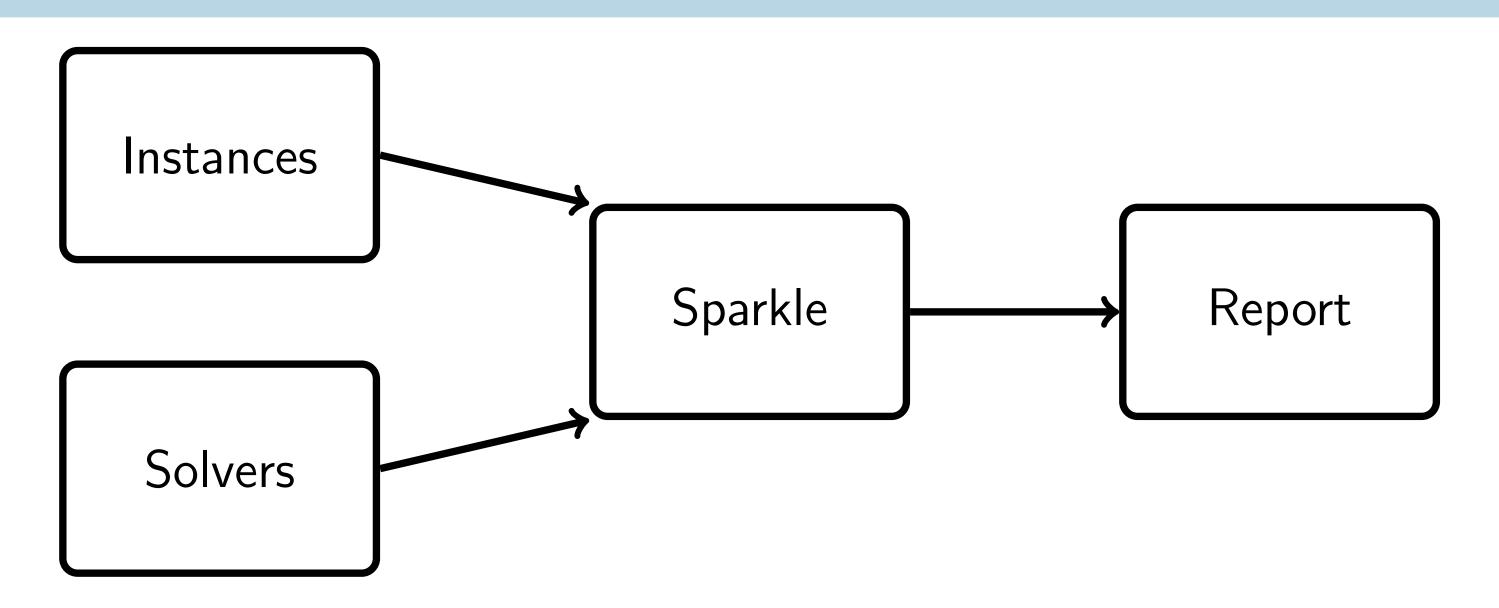
Sparkle: Towards Automated Algorithm Configuration for Everyone

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Sparkle



SIMPLE COMMANDS

- 1: Commands/initialise.py
- 2: Commands/add_instances.py -run-solver-later -run-extractor-later ../PTN/
- 3: Commands/add_solver.py -run-solver-later -deterministic 0 ../PbO-CSCCSAT-Generic/
- 4: Commands/add_feature_extractor.py -run-extractor-later ../SAT-features-competition2012/
- 5: Commands/compute_features.py
- 6: Commands/configure_solver.py -solver
 - Solvers/PbO-CSCCSAT-Generic -instance_set Instances/PTN/
- 7: Commands/generate_report_for_configuration.py

• Make meta-algorithmics more accessible

2.4 Experimental Setup

computation run is set to 8192 MB.

2.6 Marginal Contribution

amc(s) =

contribution for solver s is calculated as

vector computation on each instance is set to 32768 MB.

2.5 Constructing Portfolio-Based Algorithm Selector

achieved by leveraging the complementary strengths of the algorithms in P.

- Operate with simple commands
- Integrated best practices and avoidance of pitfalls

Feature computation: We use all the feature extractors which are presented above to compute

the feature vector for each instance. Each feature extractor will compute a feature vector for each

instance. The final feature vector is the combination of all computed feature vectors. The cutoff time

for feature vector computation on each instance is set to 90 seconds. The memory limit for feature

Performance computation: Each solver will run one time on each instance. The cutoff time

for each performance computation run is set to 10 seconds. The memory limit for each performance

Sparkle runs all the feature extractors to compute the feature vector for each instance, and store the

resulting feature data (feature vectors for all instances) in the system. Also, Sparkle runs all the solvers

to solve each instance, and store the resulting performance data in the system. After the feature-related

and the performance-related experiments are finished, by utilising the feature data and the performance

data, Sparkle uses AutoFolio [3] to automatically construct a portfolio-based algorithm selector in the

Since the primary goal of *Sparkle* is to analyse the contribution of each solver to the real state of the

the perfect portfolio selector, also known as Virtual Best Solver (VBS), and the actual portfolio

Now we give the protocol for calculating each solver's marginal contribution. Assume that we have

a set of solvers S and a portfolio selector P constructed based on a subset of S. In this report, the performance of a solver or portfolio selector is measured via the the penalised average runtime (PAR2).

which conforms with the measurement used in SAT Competitions. Let par2(P) denote the PAR2 value

Then, for each solver s, we calculate the relative marginal contribution (Rel_Margi_Contr), which

is propositional to the absolute marginal contribution, with normalisation. The relative marginal

 $rmc(s) = \frac{1}{\sum_{s' \in S} amc(s')}$

Finally, the contribution of each solver is evaluated by the relative marginal contribution

The absolute marginal contribution (Abs. Margi. Contr.) for solver s is calculated as

art, Sparkle utilises the concept of marginal contribution [4] to measure each solver's contribution to

Reporting

The System Report for Sparkle

Automatically generated by *Sparkle* (version: Sparkle_SAT_Challenge_2018) 1st April 2019

100 110

1 Introduction

Sparkle [2] is a multi-agent problem-solving platform based on Programming by Optimisation (PbO) [1], and would provide a number of effective algorithm optimisation techniques (such as automated algorithm configuration, portfolio-based algorithm selection, etc) to accelerate the existing solvers. This experimental report is automatically generated by Sparkle. This report presents the system status for Sparkle.

2 Experimental Preliminaries

In this section, we present the experimental preliminaries, including the list of solvers, the list of feature extractors, the list of instance classes, the information about experimental setup and the information about how to construct a portfolio-based algorithm selector in *Sparkle*.

2.1 Solvers

There are 3 solver(s) submitted in Sparkle, and the list of solver(s) is given as follows.

1. CSCCSat_wrapper_sparkle

2. Lingeling_wrapper_sparkle 3. MiniSAT wrapper sparkle

2.2 Feature Extractors

There are 1 feature extractor(s) submitted in *Sparkle*, and the list of feature extractor(s) is given as follow.

There are 10 instance(s) submitted in *Sparkle*. All instance(s) are classified into 1 instance class(es)

 $1. {\bf SAT}\-features\-competition 2012_revised_without_SatELite_sparkle$

2.3 Instance Classes

1. Sparkle test instances, number of instances: 10

and the list of instance class(es) is given as follows.

3 Experimental Results

selector.

In this section, the related experimental results in *Sparkle* are presented and analysed.

• PbO-CCSAT-Generic (configured), PAR10: 8.500000

• PbO-CCSAT-Generic (default), PAR10: 287.646364

The empirical comparison between the PbO-CCSAT-Generic (configured) and PbO-CCSAT-Generic (default) on the training set of PTN is presented in Figure 2.

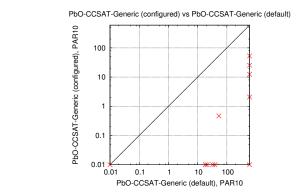


Figure 2: Empirical comparison between the PbO-CCSAT-Generic (configured) and PbO-CCSAT-Generic (default) on the training set of PTN.

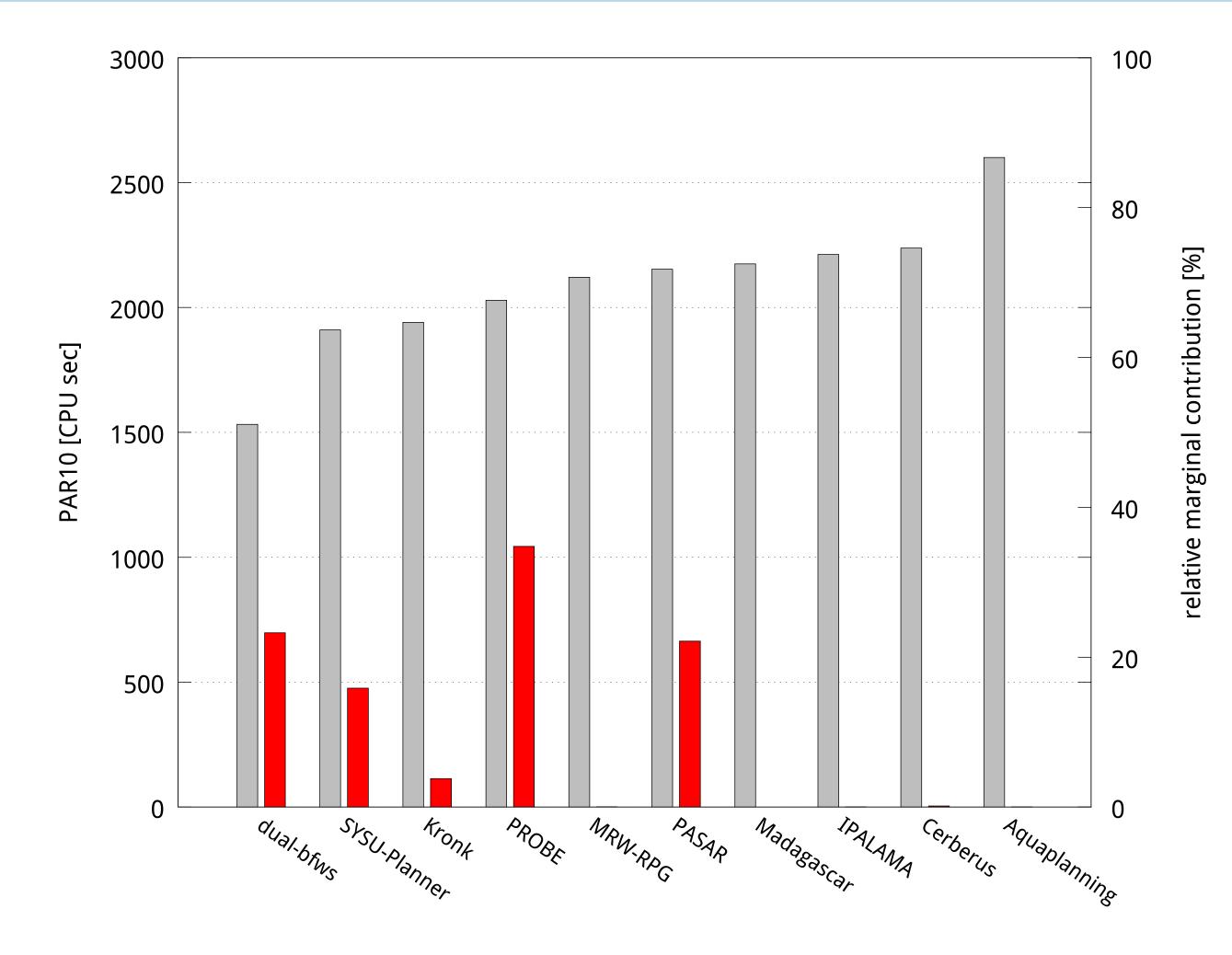
References

 $\label{eq:communication} [1] \mbox{ Holger H. Hoos. Programming by optimization. Communications of the ACM, 55(2):70-80, 2012.$

[2] Holger H. Hoos. Sparkle: A pbo-based multi-agent problem-solving platform. Technical report, Department of Computer Science, University of British Columbia, 2015.

[3] Frank Hutter, Holger H. Hoos, and Kevin Leyton-Brown. Sequential model-based optimization for general algorithm configuration. In *Proceedings of the 5th International Conference on Learning* and Intelligent Optimization (LION 5), pages 507–523, 2011.





1 2 3

- Contribution per solver Cite configurator
- Cite selector

• Process description

References

- [1] Shapley, L. S. (1953). A value for n-person games. In Contributions to the Theory of Games, volume II, pages 307âĂŞ317. Princeton University Press.
- [2] Xu, L., Hutter, F., Hoos, H. H., and Leyton-Brown, K. (2012). Evaluating component solver contributions to portfolio-based algorithm selectors. In Proceedings of the 15th International Conference on Theory and Applications of Satisfiability Testing (SAT 2012), LNCS 7317, pages 228âĂŞ241.

• Marginal contribution [Xu et al 2012]: How valuable is this solver to the selector

FUTURE

- Further simplifications, such as inferring the parameter space
- Shapley value [Shapley 1953]?
- Your ideas?