

Per-class AS for black-box optimisation

AS for BBO scenarios without feature computation

by selecting based on a priori known features.

A Per-Class Automated Algorithm Selection Framework for General Black-Box Optimisation

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Summary

Main idea

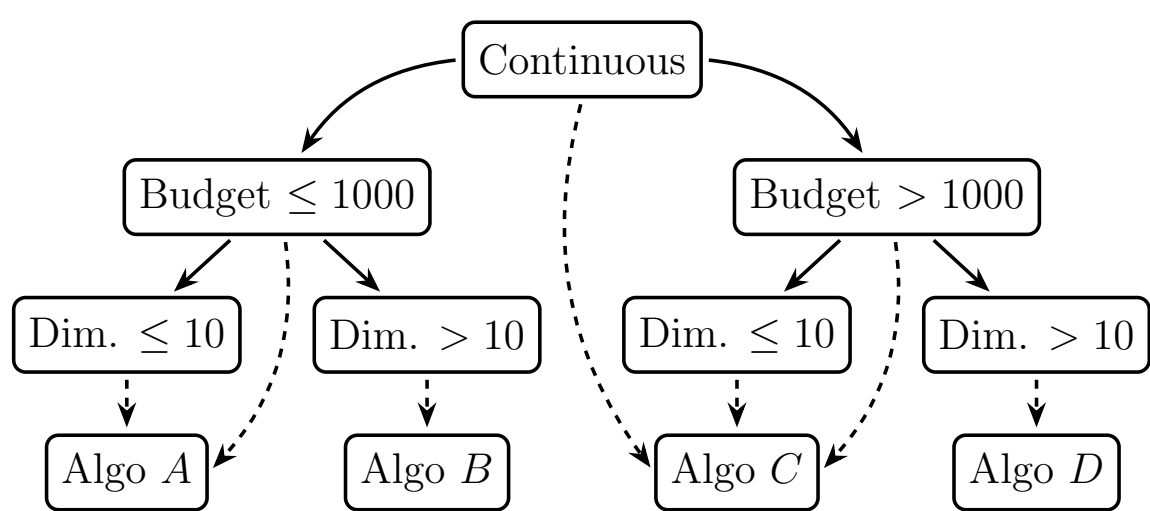
- **Instance-specific features:** Too costly, not always available
- **Per-class AS:** Only use a priori known features
- **BBO known features:** Variable type, evaluation budget, ...
- **Group instances in classes:** More specific than the SBS

Results

- Definition of per-class algorithm selection
- Framework for principled and reproducible PCAS construction
- PCAS more accurate and specific than hand designed selectors

Benefits

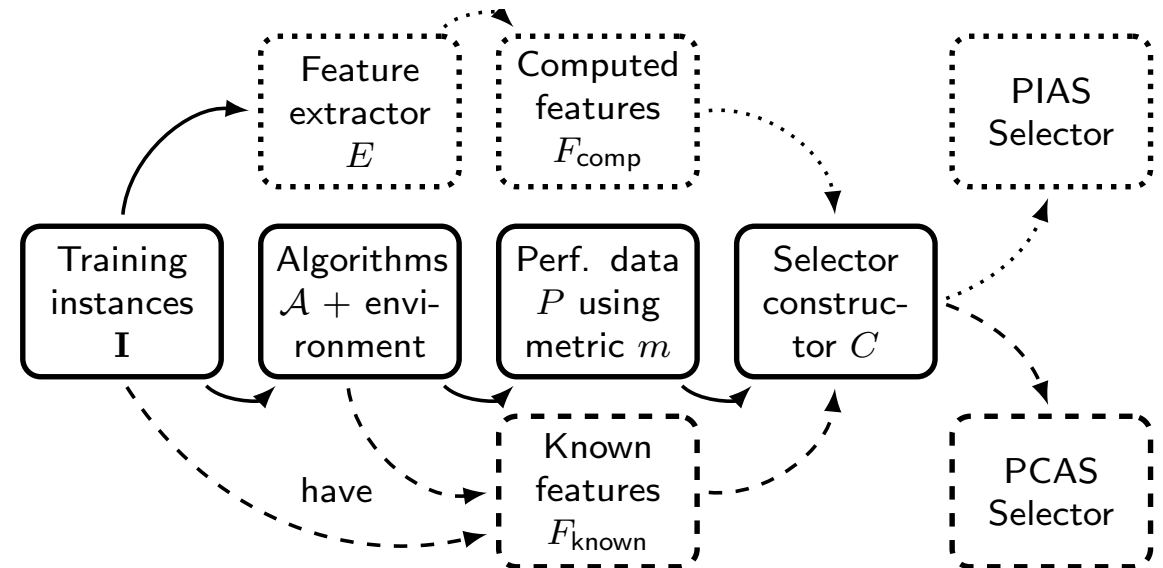
- Quick access to a good initial algorithm for any BBO problem
- AS possible for BBO when no feature computation possible
- Easily integrate grey-box features when available
- Facilitate BBO benchmarking view at each class level
- Possible fallback for PIAS systems when out of distribution



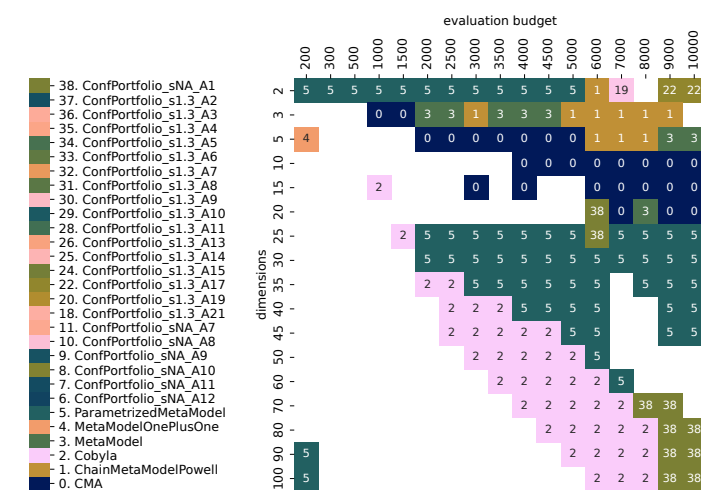
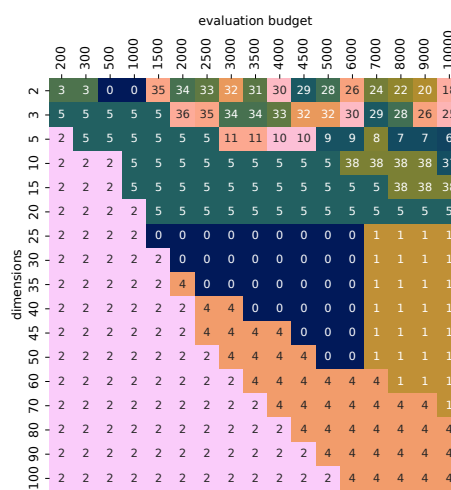
Split instances into classes and select at available detail level

Example setup

- **Training:** 24 BBOB functions, 1 instance, 25 runs
- **Testing:** Distinct BBOB instances, MA-BBOB functions
- **Dimensionality:** $d \in [2, 100]$
- **Budgets:** 200 to 10 000 evaluations
- **Algorithms:** 34 used by NGopt39 (Nevergrad) in same domain

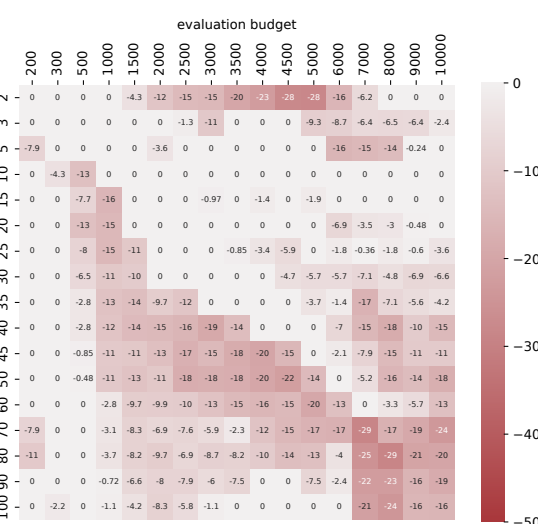
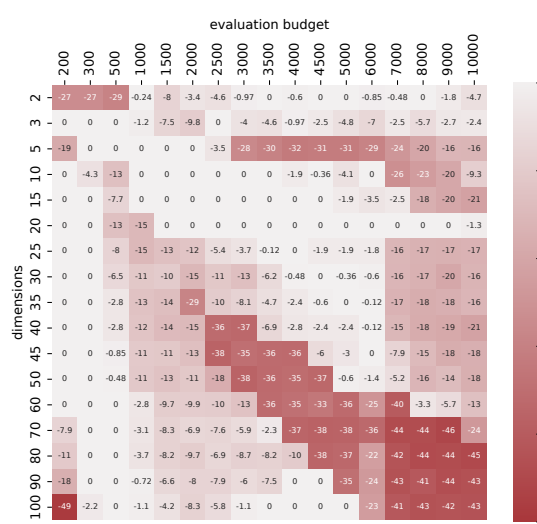


Comparison between PIAS (dotted) and PCAS (dashed) components



Experiments

- Select by ranking based on number of wins, top- n results
- Compare percentage of instances with non-zero log loss to VBS



Next steps

- Extend to
 - Problems with constraints, multiple objectives, ...
 - Grey-box features
 - Other performance measures
- Incorporate more problems/instances and algorithms
- Efficiently choose which algorithms to run on which instances