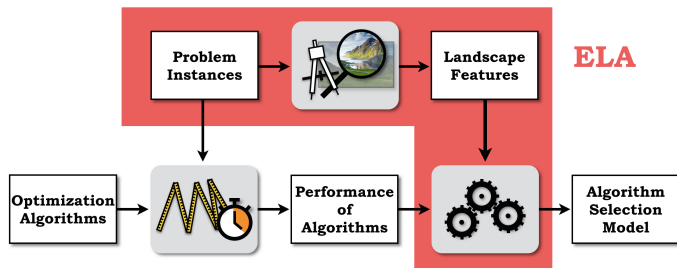


General Idea of Exploratory Landscape Analysis

Introduction

Goal:

- improve understanding of (continuous black-box) problems
- describe relationship between algorithm behavior and underlying problem
- ultimate goal for algorithm selection problem¹ (ASP): select the “best” algorithm



¹Rice, J. (1976). *The Algorithm Selection Problem*. In: *Advances in Computers* (pp. 65 – 118).

Idea of *Exploratory Landscape Analysis (ELA)*:

- characterize black-box problems by numerical (and thus automatically computable) values
- start with very simple features without clear purpose
- match existing high-level features² with our ELA features

²high-level features = properties / characteristics of the problem landscape as categorized by an expert

Introduction

Notes I:

- functional relationships are unknown when designing features (usually one has a vague idea of what kind of property one would like to “measure”)
- pure numbers of a single feature on a single problem are basically meaningless
 - ↪ look at combination of features and/or compare the values across problems

Notes II:

- try to match the features to high-level characteristics⁵ (multimodality, funnel structure, etc.) of optimization problems
- this enables recognizing important problem properties quickly (and without consulting an expert)

⁵usually via classification models, whose “class labels” are the problem properties

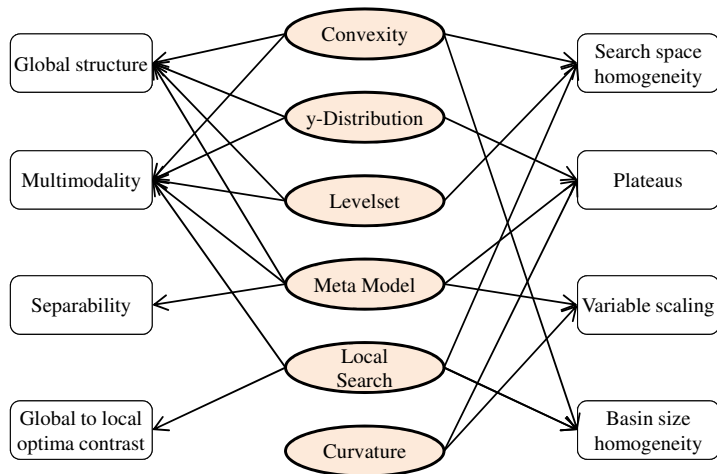
Notes III:

- features are based on initial design of samples x_{i1}, \dots, x_{iD} and their corresponding fitness values y_i , $i = 1, \dots, n$
- given an evaluated initial design⁶, most ELA features are for free
↪ they don't need any further function evaluations
- multiple different feature sets already exist, and we will introduce some of them on the following slides⁷

⁶usually a well-spread sample (LHS, random uniform sample, etc.); however, using the initial population of an optimizer is also possible

⁷for further details, please attend “ELA Tutorial” at PPSN 2018 ;-)

Introduction



Mersmann, O., Preuss, M. & Trautmann, H. (2010). *Benchmarking Evolutionary Algorithms: Towards Exploratory Landscape Analysis*. In: Proceedings of PPSN XI (pp. 71 - 80).

Notes I:

- flacco: **F**eature-Based **L**andscape **A**nalysis of **C**ontinuous and **C**onstraint **O**ptimization Problems
- unified interface for multiple (single-objective) sets of configurable features
- stable release on CRAN / developers version on GitHub
- multiple vizualisation techniques (partially shown on these slides)

Notes II:

- flacco also comes with a platform-independent web-application

Feature	Value
ela_meta.lin_simple.adj_r2	0.08
ela_meta.lin_simple.intercept	-355.35
ela_meta.lin_simple.coef.min	0.79
ela_meta.lin_simple.coef.max	0.94
ela_meta.lin_simple.coef.max_by_min	1.20
ela_meta.lin_w_interactadj_r2	0.18
ela_meta.quad_simple.adj_r2	0.07
ela_meta.quad_simple.cond	1.71
ela_meta.quad_w_interactadj_r2	0.23
ela_meta.costs_fun_evals	0.00
ela_meta.costs_runtime	0.01

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⁸Link to GUI: <https://flacco.shinyapps.io/flacco/>

Notes III:

- tracks # of function evaluations and run time - per feature set
- FLACCO is described in our CEC paper:
Kerschke, P. & Trautmann, H. (2016). *The R-Package FLACCO for Exploratory Landscape Analysis with Applications to Multi-Objective Optimization Problems*. In: Proceedings of CEC 2016.
- further information on FLACCO, its GUI, or the contained feature sets can be found here:
Kerschke, P. (2017). *Comprehensive Feature-Based Landscape Analysis of Continuous and Constrained Optimization Problems Using the R-Package flacco*.
In: <https://arxiv.org/abs/1708.05258>.