

# **Fine-grained Algorithm Design and Engineering**

Solver 101 08.11.2024





#### **Overview**

- why we do algorithmic challenges
  - how challenges work
- what you will learn
- how to structure a solver
- test instances
- organization

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#### Why Algorithmic Challenges?

theoretical research focusses on worst case analysis

- or special cases
- in practice, problems are solvable efficiently
  - even without special cases

How well do theoretic approaches work? Can we find new ones?

- better understand engineers
- baseline: beat general purpose solvers!





at least more efficiently than suspected





# **Challenge Framework**



- differences based on specific challenge
- around 10 to 100 participating teams
- challenge announcement, publication of partial test set
- solver submission until deadline, running scoreboard
- evaluation on (additional) private instances
  - similar to public instances
    might get you disqualified
  - scoring based on feasibility and optimization criterion
  - sometimes: points for solver idea
- winning solvers: invitation to associated conference?

# In this course, you will ...



- ... work on more real-world problems
- ... learn approaches to tackle general algorithmic problems
- ... differentiate between theoretic results and practical possibilites
- ... engineer a solver from scratch
- ... apply standard algorithmic techniques and learn about their limitations ...
- ... use heuristics and general purpose solvers
- ... learn to evaluate algorithmic approaches
- employers like algorithmic challenges

# Solver 101





may use existing solvers

Disclaimer: Non-exhaustive, personal preference, only basic ideas



have your own reasonable internal format

input/output might change depending on source format, visualization, etc.



#### Where the Magic Happens



Example: Vertex Cover

• Want specific material? Ask us!







Param. Algos Structural Restrictions

see param. alg. lecture

https://scale.iti.kit.edu/teaching/2024ws/param\_algo/

- kernelization
- keep track of your parameters

Classic Algo. Approaches

runtime exponential in parameter, polynomial in input size

structural parameter (e.g. treewidth, planarity), output size

• parameter small  $\implies$  fast runtime

General Purpose Solvers

General Tools

- reduce optimization to decision problem
- selected techniques:
  - branch & bound not only in parameterized setting!
  - treewidth-DP
  - ILP with few variables  $\implies$  reduction to ILP solver



not in solver,

LP-solving in P!

create heuristics

ok for evaluation

smaller instances fast

kernelization

Classic Algo. Approaches

> General Purpose Solvers

General Tools Iterative / Dynamic Application

Heuristics

EvalMaxSat (github: FlorentAvellaneda)

FindMinHS (github: Felerius, i.a. KIT dev-ed)

GLPK (GNU Linear Programming Kit)

- NP-membership gives reduction to SAT, (M)ILP
  - ridiculously strongly optimized solvers
  - baseline for all practical use cases
- ask for partial solution, dynamically add constraints
- Decide what to do? Educated guesses with heuristics!
   might speed-up computation
- greedy and local search often helpful
- approximate the result
  - prune your search-tree!





Composition Distinguisher Verification fast verification allows better use of heuristics kernelization many instances special enough to be easily solvable what kind of special? Classic Algo. Approaches distinguishing itself complex, keep track of parameters degree distribution, guess (& bound) your parameters General runtime depending on interaction, need to iterate procedure Purpose Solvers example: strong reduction rules, dynamic reduction to ex. solvers, heuristically making decisions for next iteration General circumvent worst of worst cases: add randomness Tools alternative formulation of the problem?

#### **Test Instances**



- look at previous write-ups
  - public test instances
  - real-world instances
    SNAP project (Stanford Large Network Dataset Collection)
  - randomly generated instances (often hard to solve)
  - other instances harder to predict (and generate)
- course-specific: instances encoding other problems (fine-grained) reductions?
- persistent output  $\implies$  file output
  - evaluate your solutions  $\implies$  feasible?
- visualize your instances! e.g. graphviz
- scripts to run/evaluate large batches

#### **Miscellaneous**

start small: have your boilerplate code running

- input/output, validation, testing (metrics or visuals)
- focus on easy instances: solve them fast
  - allow unsolved outputs, look at unsolved parts
  - collect hard instances, investigate them
  - find approaches to next portion of instances
- switch to general solver should not happen too early
  - often tradeoff between simplicity and generality
  - interesting insights often found for specific instances, generalized later
- look at existing algorithmic ideas first

Be Curious!





original by CGP Grey

# Organizational



repository, please invite us

use suitable libraries!

- all code, test instances and output, visualizations, etc.
- scripts for building, evaluation, ...

readme

- explain usage: how to build and run solver, how to evaluate
- contribution (bullet points)
  - what was taken from which source, who worked on what part

tbd: How To Report in December (around 15.12) – announcement via email

# Q&A now!