



# Visualizations to Explain Algorithm Performance

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## Context and Contributions

**Context:** Constraint Programming (CP) is powerful paradigm for modeling and solving complex decision problems in management and engineering, including resource allocation, scheduling, product configuration, and design.

These problems are represented as Constraint Satisfaction Problems (CSPs), which are NP-complete, making their solving cost difficult to predict.

**Contributions:** We design visualizations that summarize the behavior of the CP solver to

1. Explain the cost of the solver
2. Allow the users to design new algorithms and heuristics

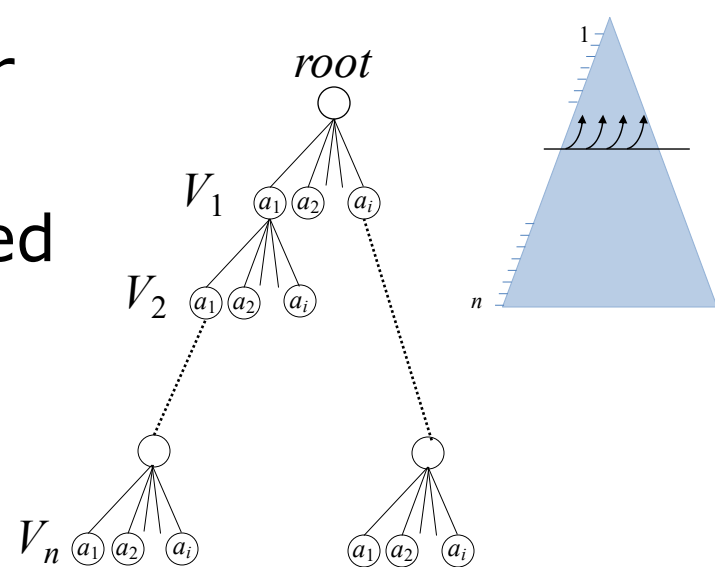
## Solving CSPs with Search

**Constraint Satisfaction Problem:** Given

- A set of *decisions* (variables), *options* for each decision (values), and
- A set of *constraints* restricting the allowed combinations of options to decisions (combinations of values to variables),

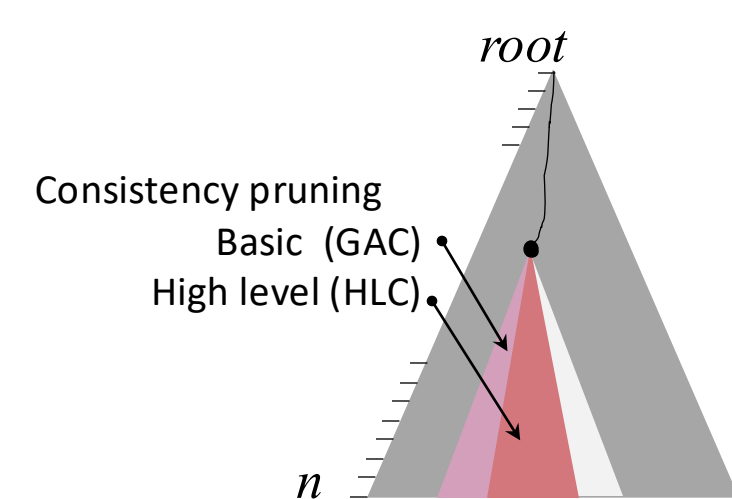
Determine if the problem has a solution that satisfies all constraints

Example: a Sudoku puzzle



### Algorithms

1. *Backtrack search:* Enumerates all combinations of values for variables
2. *Ordering heuristics:* Choose which variable to instantiate first
3. *Consistency algorithms:* Prune inconsistent combinations



## Parameters and Visualizations

### Cost of search

1. Number of variable instantiations: #NV
2. Number of backtracks: #BT
3. Number of calls to a consistency algorithm
4. CPU time

### Visualizations

1. BpD: Number of backtracks per depth
2. VIpD: Number of variable instantiations per depth
3. CpD: Number of calls to a consistency algorithm per depth
  - #Wipeouts: Consistency prunes the entire subtree (**super effective**)
  - #Filter: Consistency reduces size of subtree (**beneficial**)
  - #No-filter: Consistency does not affect subtree (**wasted efforts**)

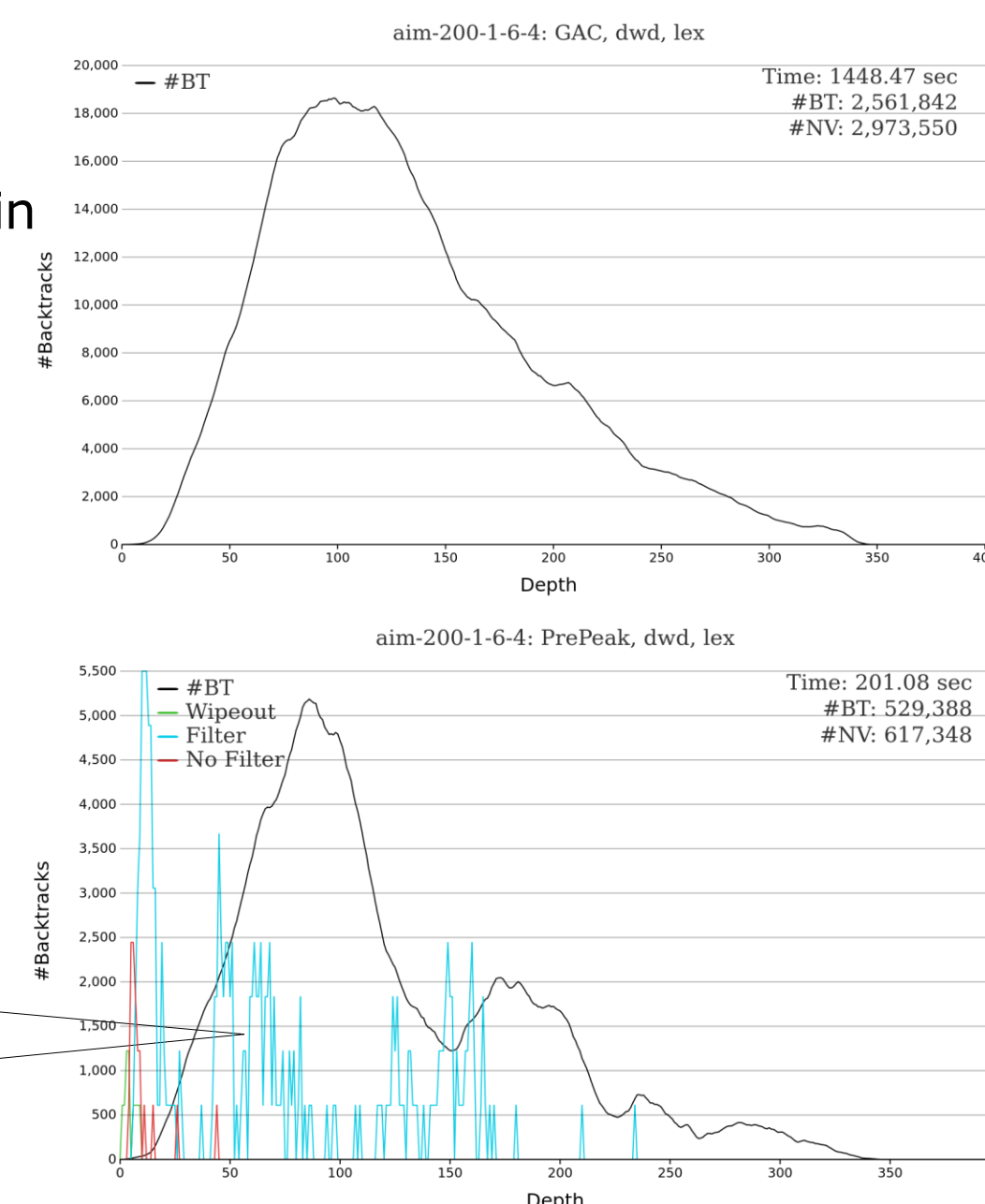
## High-Level Consistency

### Solving aim-200-1-6-4

- GAC, the basic consistency algorithm, takes more than 24 min
- POAC, a high-level consistency applied reactively because it is expensive, takes less than 4 min

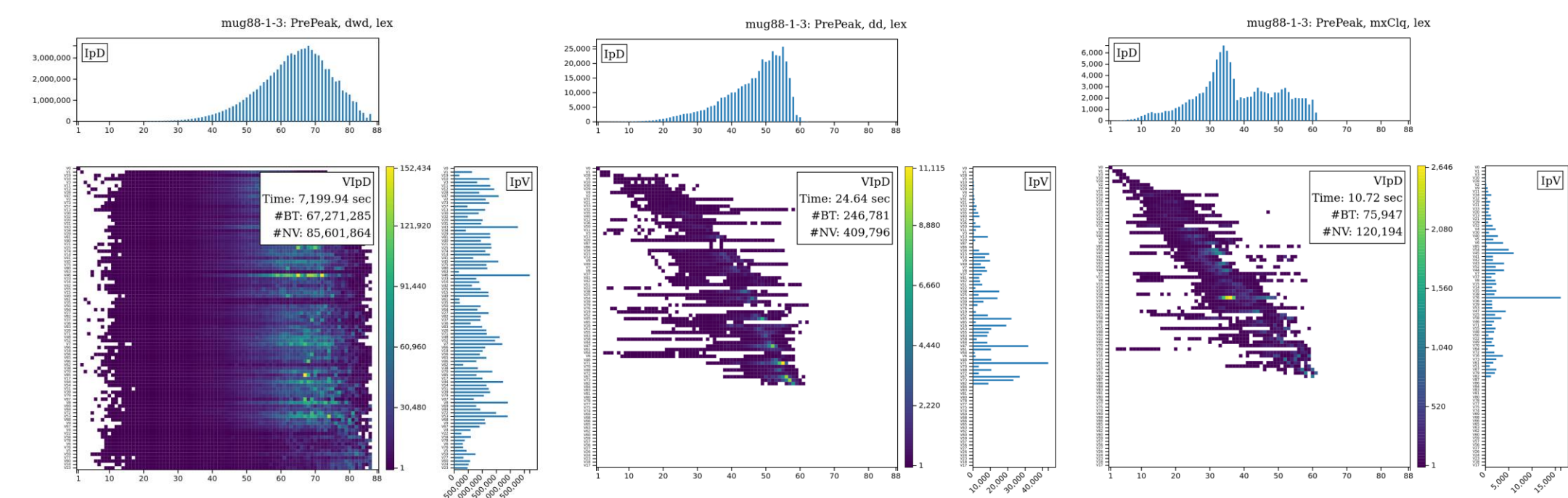
Consistency	Basic GAC	HLC, reactively POAC
CPU time	1,448.5 sec	201.08 sec
#BT	2,561,842	529,388
#NV	2,973,550	617,348
max #BT	18,632	5,185

- #Calls to (expensive) HLC
- Wipeouts: **super effective**
  - Filter: **beneficial**
  - No filter: **wasted efforts**



## Ordering Heuristics

### mug88-1-3 (UNSAT)

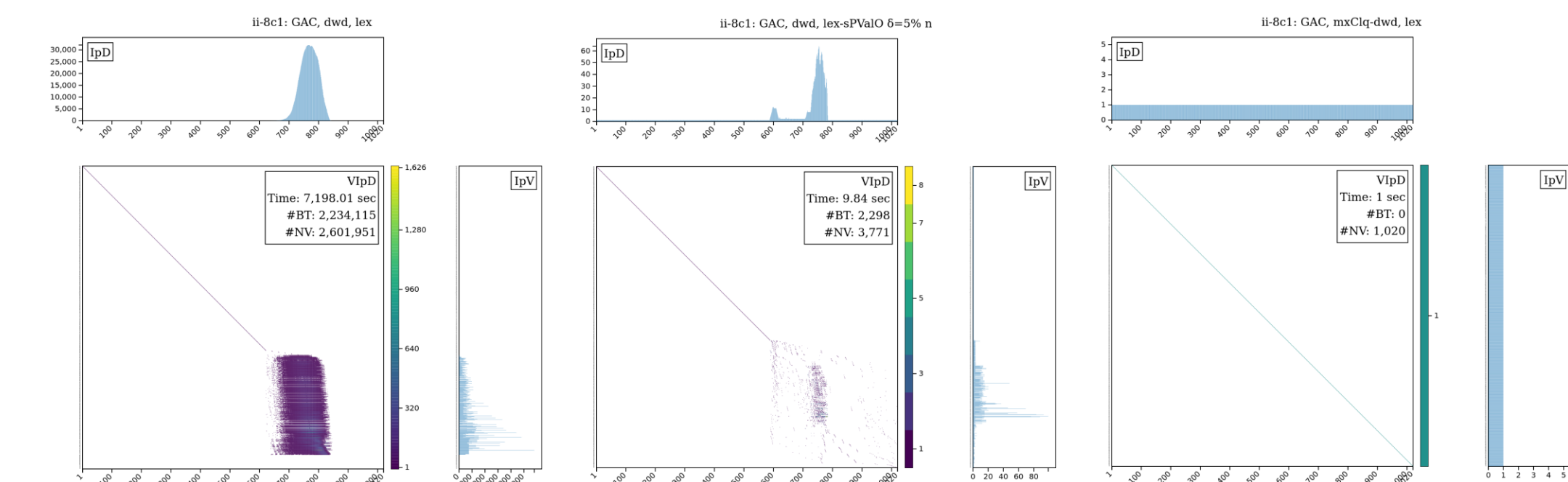


dwd, currently the leading dynamic variable ord. heuristic, is sometime erratic

dd, an older dynamic variable ord. heuristic, is sometimes more stable than dwd

MaxClique, our static variable ord. heuristic, exploits the problem's structure and performs best

### ii-8c1 (SAT)



dwd, again, is erratic: search does not terminate within two hours

Reactively applying PVal0 [Geelen 92], a costly value ord. heuristic, terminates in less than 10 seconds

MaxClique, our static variable ord. heuristic, exploits the structure of the CSP and solves this instance *backtrack-free!*

## Future Work

- Visualize the constraint graph
- Coordinate various visualizations
- Generate animations of significant search regimes

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