
Visualizations to Summarize Search Behavior

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Introduction

- Backtrack search when solving a constraint satisfaction problem (CSP) suffers from thrashing
- Enforcing a consistency property reduces thrashing at the cost of further processing
- The tradeoff of running a higher-level consistency (HLC) is poorly understood.
- We propose to summarize search

Background

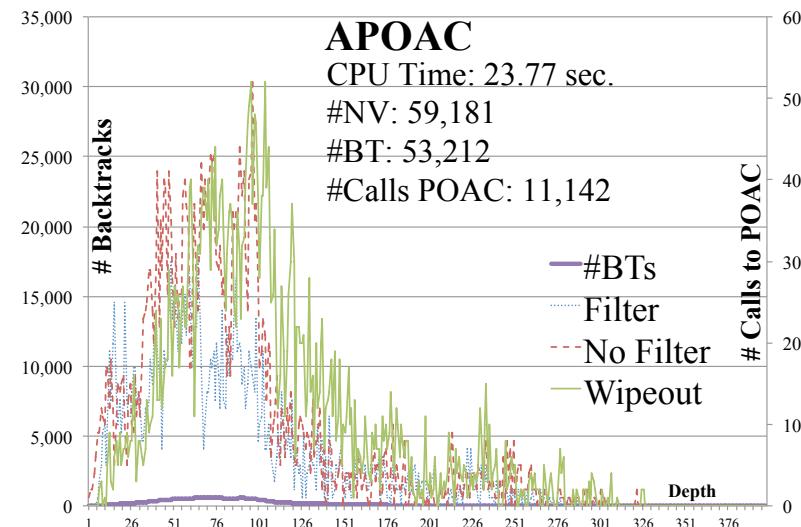
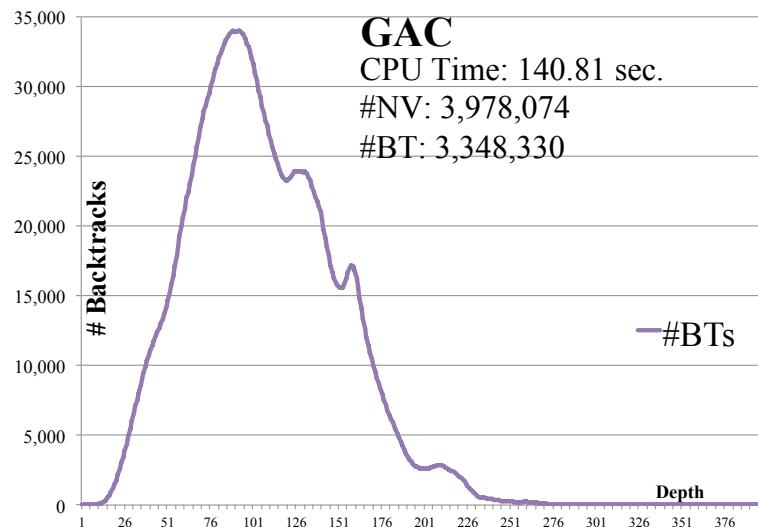
- A constraint satisfaction problem (CSP) consists of a set of variables, the variables' domains, and constraints.
- A solution is a domain to variable assignment that satisfies all constraints.
- Backtrack search is only sound and complete method of solving a CSP.

Background

- Thrashing is the main malady of search
 - Repeatedly performing same logic
- Applying a consistency algorithm after variable instantiation reduces thrashing
- Past visualizations focus on debugging and inspecting search
 - Detect thrashing through isomorphic subtrees
 - Debug individual issues by investigating the state of variables, propagators, etc.

Background

- Woodward et al. propose to measure:
 - thrashing with Backtracks per Depth (BpD)
 - HLC cost with consistency calls per depth (CpD)



Contributions

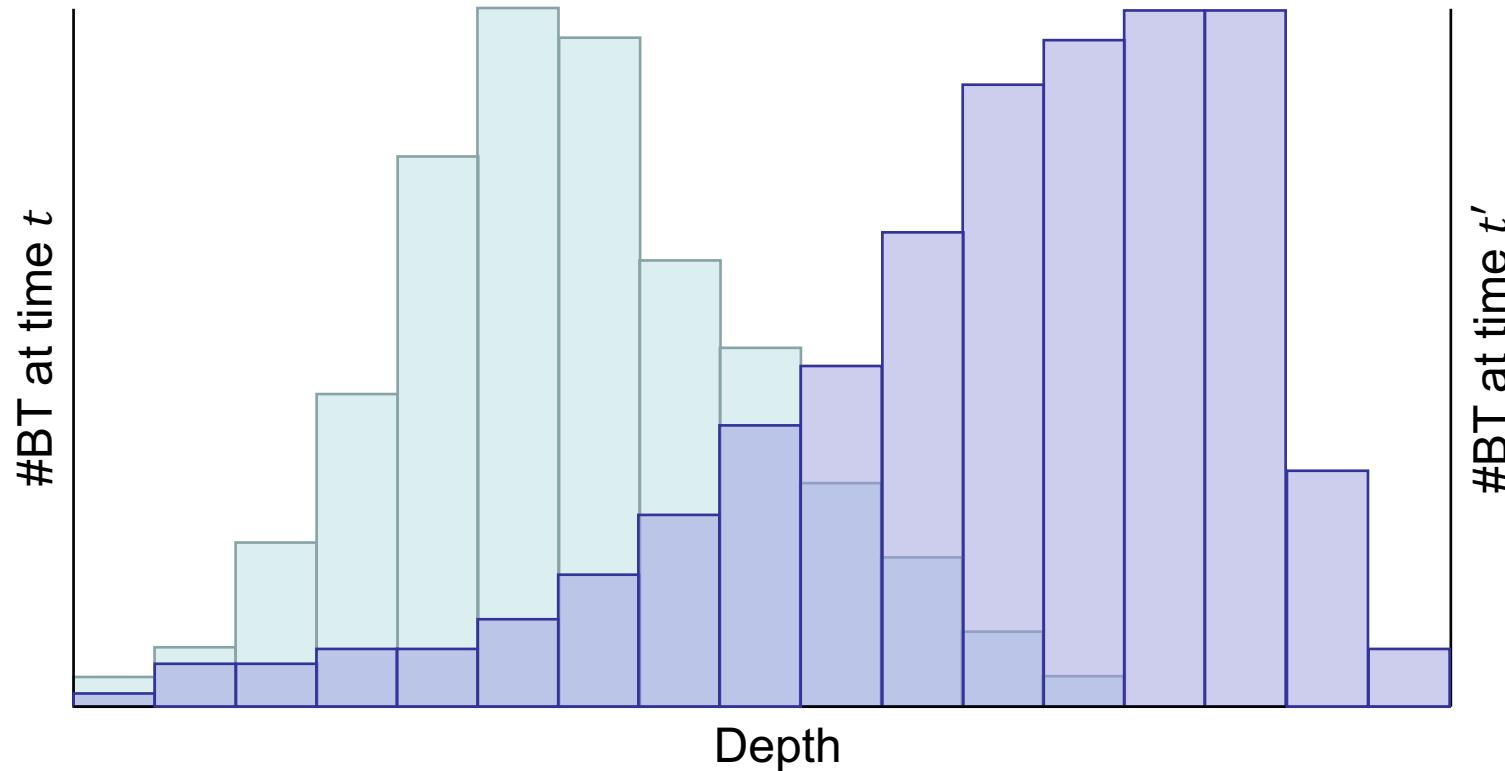
1. Criteria for computing distance between two time samples based on the BpD
2. A clustering technique for summarizing search into a history of qualitatively distinct regimes
3. A new visualization that examines the behavior of variable ordering heuristics

Outline

1. Introduction
2. Background
3. Contributions
4. **Outline**
5. Distance between timestamps
6. Summarizing search history
7. Example: Analyzing structure
8. Visualizing variable ordering
9. Example: Variable ordering
10. Conclusions and discussion

Distance between timestamps

- Design goal: capture the “shape” change



Distance between timestamps

- Adapt Kulbak-Liebler Divergence
- BpD distribution:
 - Use additive smoothing to account for zero-probability intervals

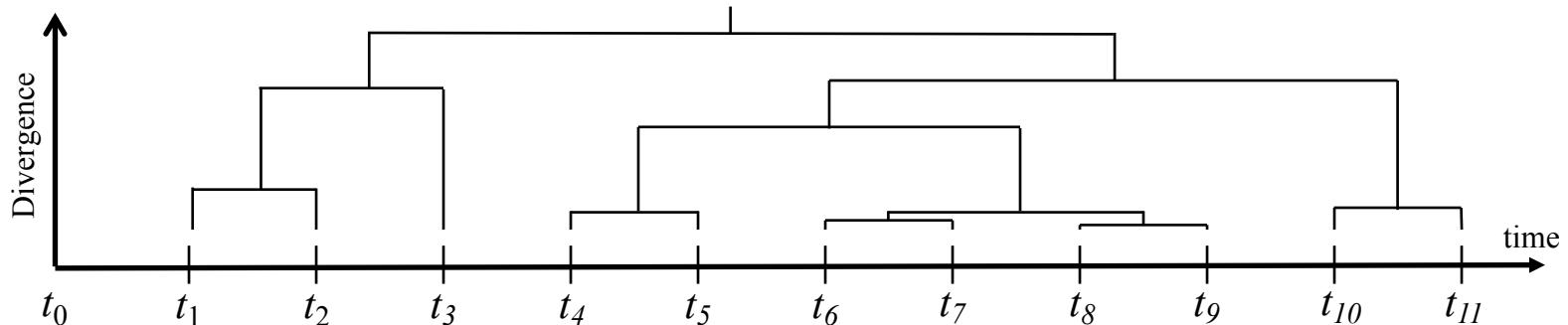
$$\hat{p}(d, t) = \frac{\text{BT}(d, t) + \alpha}{\sum_{d \in D} \text{BT}(d, t) + \alpha d_{\max}}$$

Distance between timestamps

$$\text{div}(t, t') = \max \left(\sum_{d \in D} \widehat{p}(d, t) \log \left(\frac{\widehat{p}(d, t)}{\widehat{p}(d, t')} \right), \right.$$
$$KL(t, t') \xrightarrow{\quad} \left. \sum_{d \in D} \widehat{p}(d, t') \log \left(\frac{\widehat{p}(d, t')}{\widehat{p}(d, t)} \right) \right)$$
$$KL(t', t) \xrightarrow{\quad} \left. \sum_{d \in D} \widehat{p}(d, t) \log \left(\frac{\widehat{p}(d, t)}{\widehat{p}(d, t')} \right) \right)$$

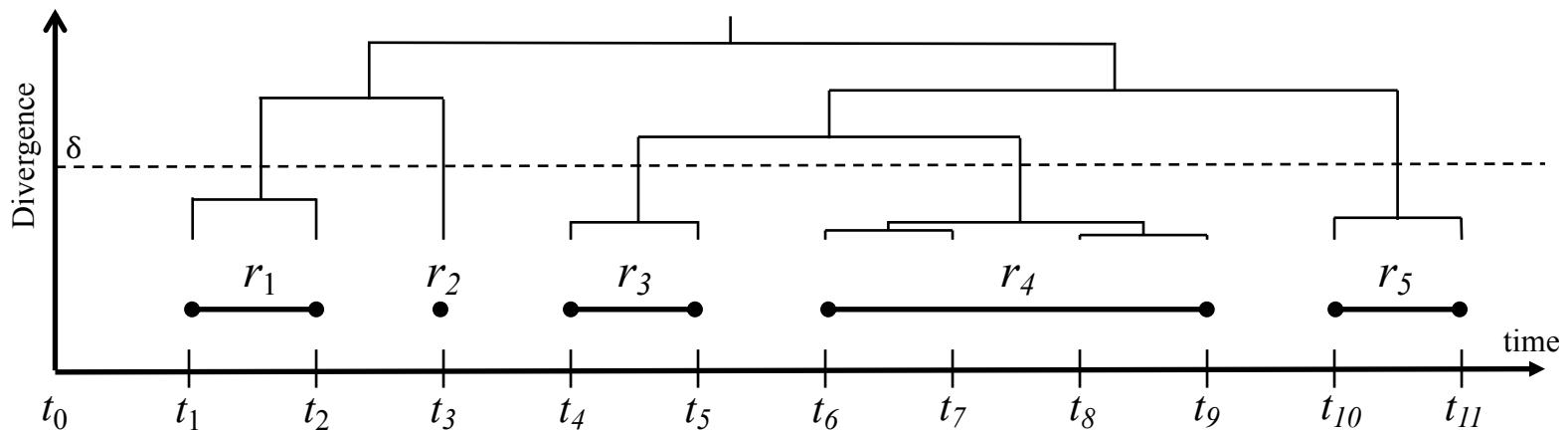
Summarizing search history

- Create a clustering tree
 - Use agglomerative, hierarchical clustering
 - Only merge temporally adjacent clusters
 - Each cluster's representative is the middle of the interval of included timestamps

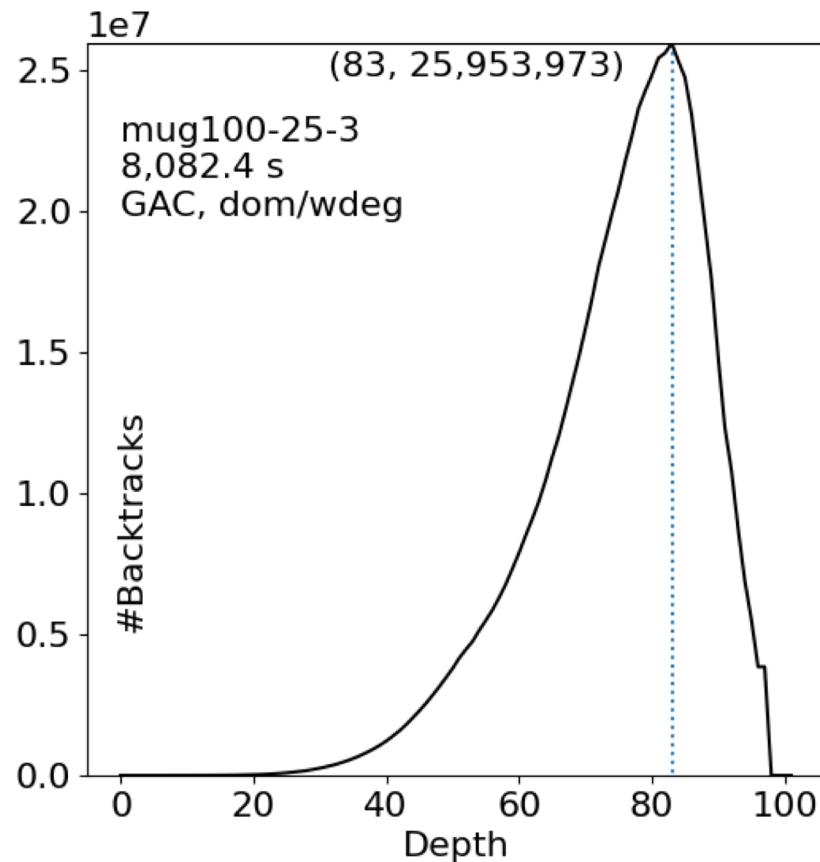


Summarizing search history

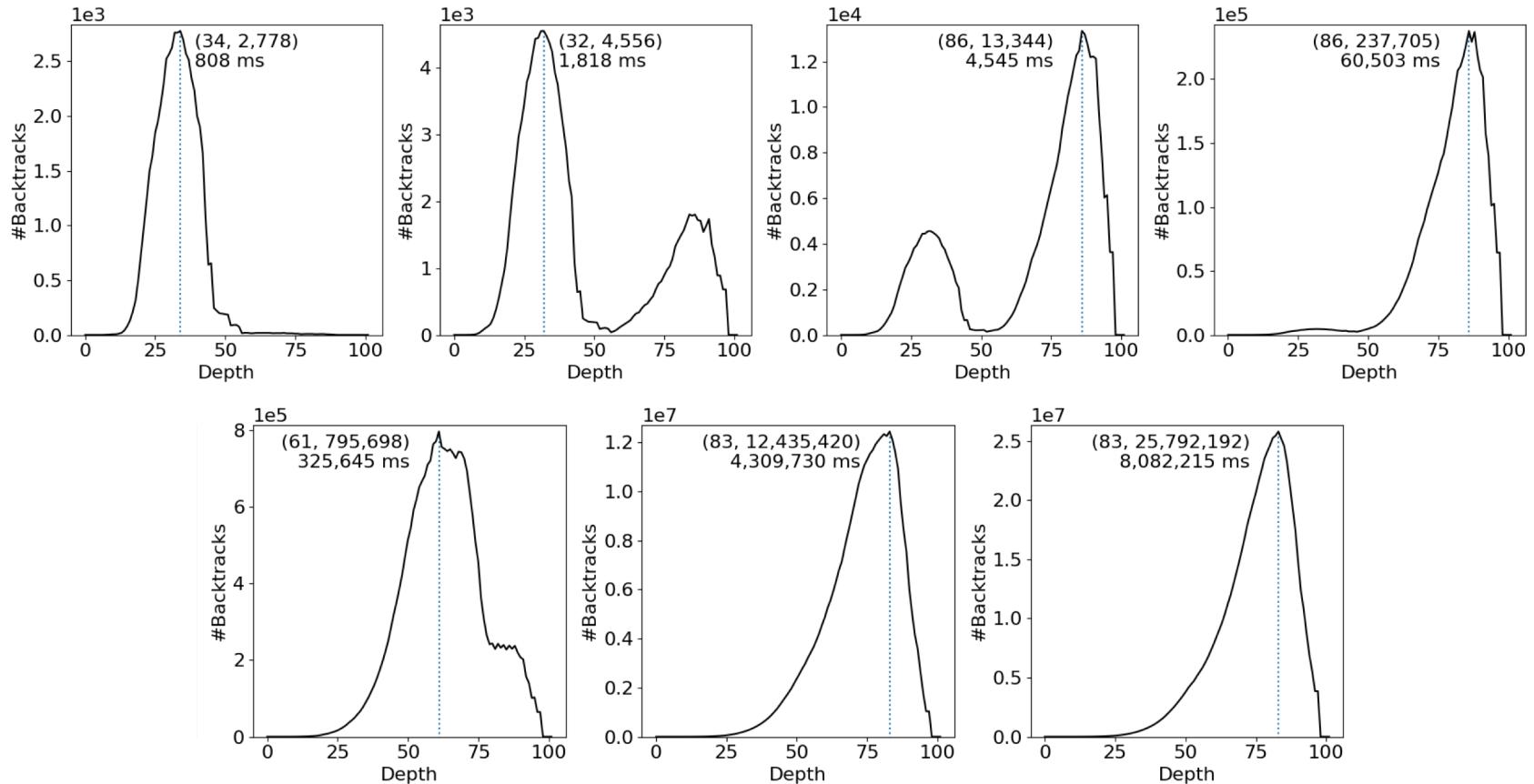
- Create a summarization tree
 - Cut the tree at a user-defined δ or
 - Cut the tree to include a n regimes



Example: Analyzing structure

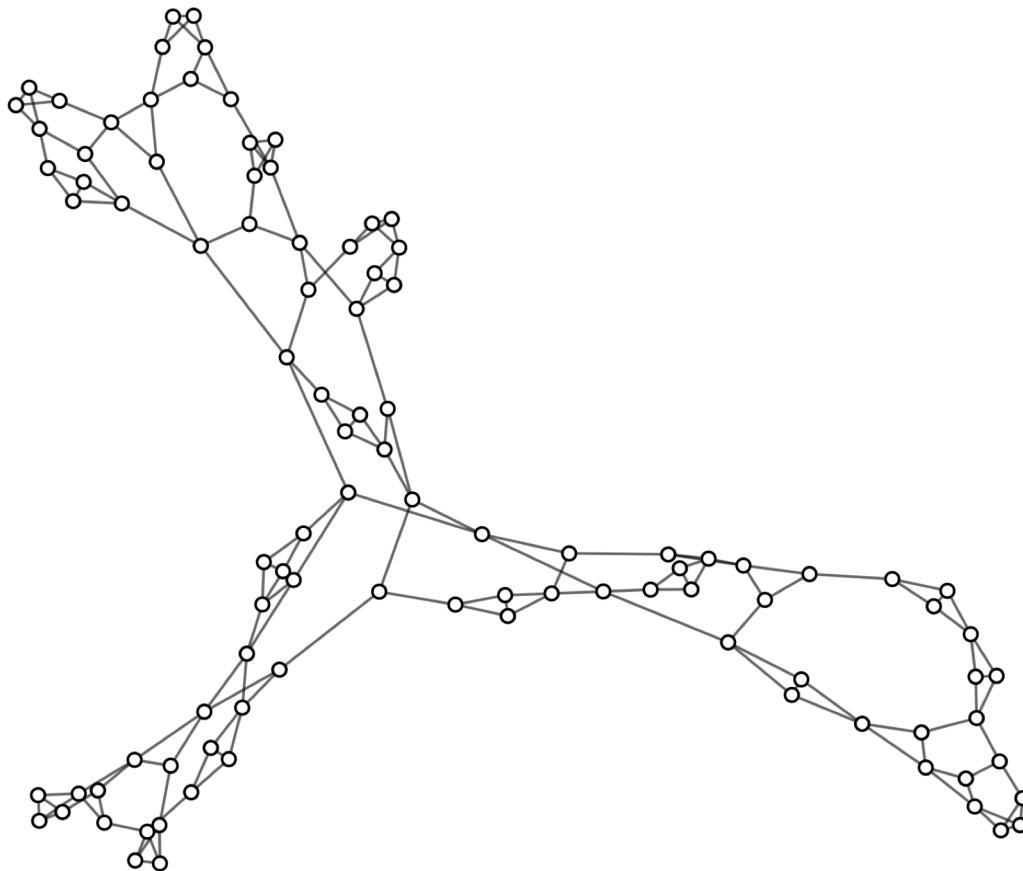


Example: Analyzing structure



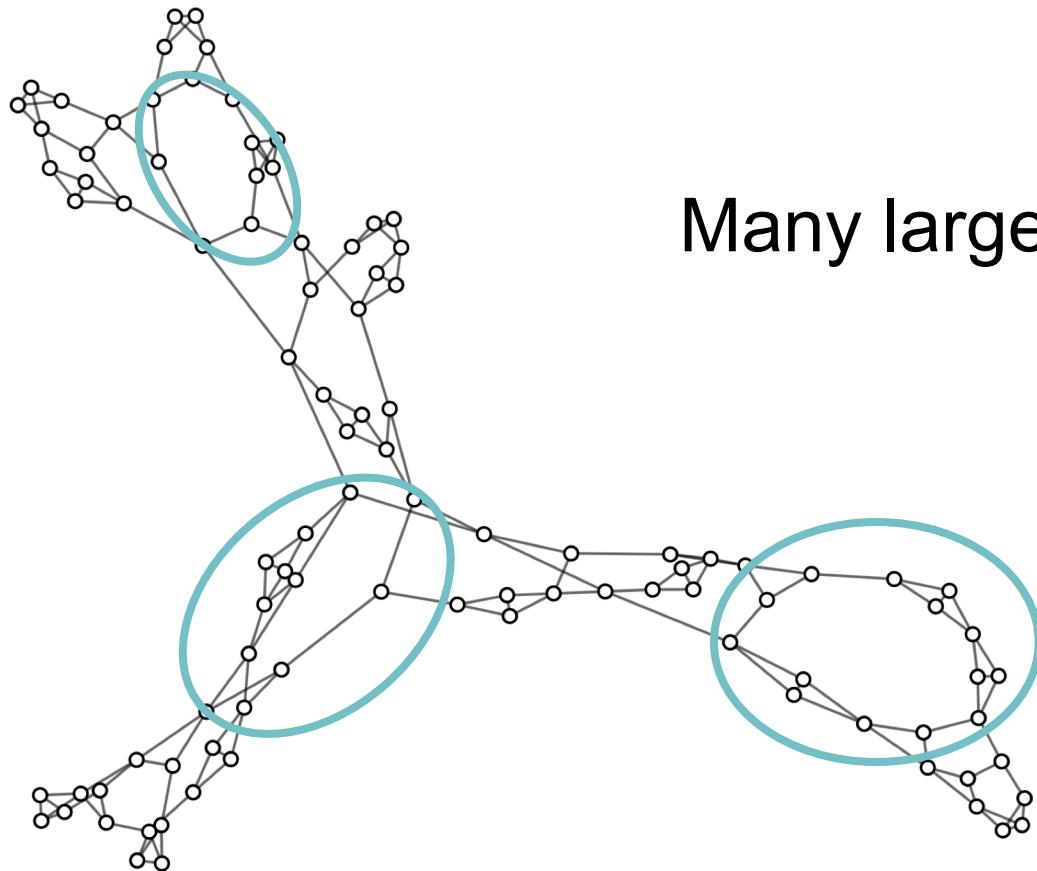
0.1-summarization of GAC on MUG100-25-3

Example: Analyzing structure



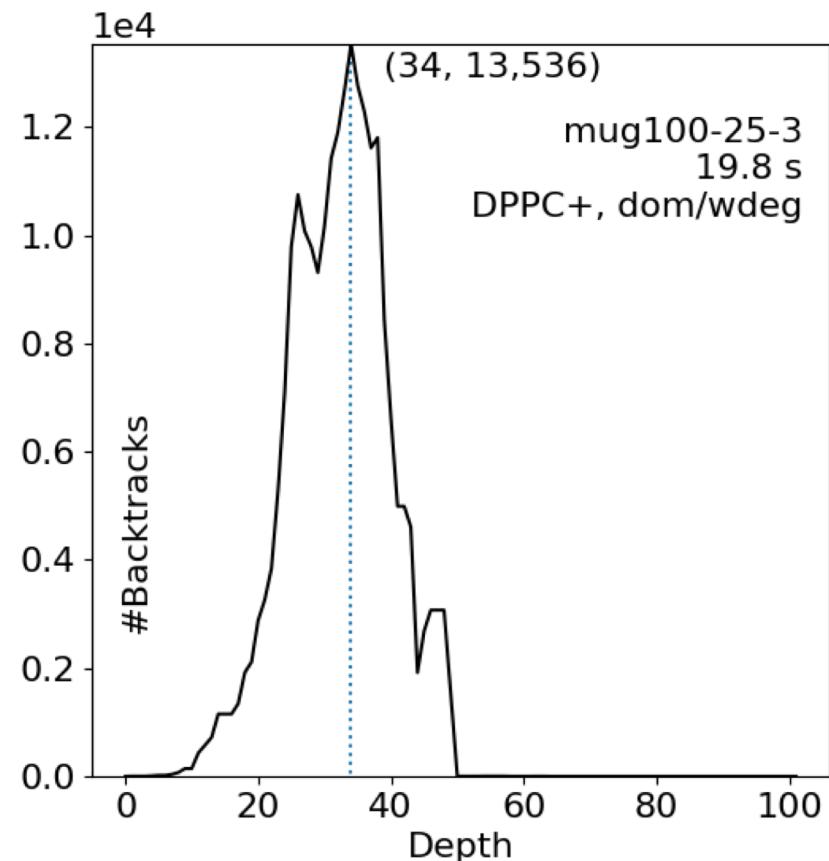
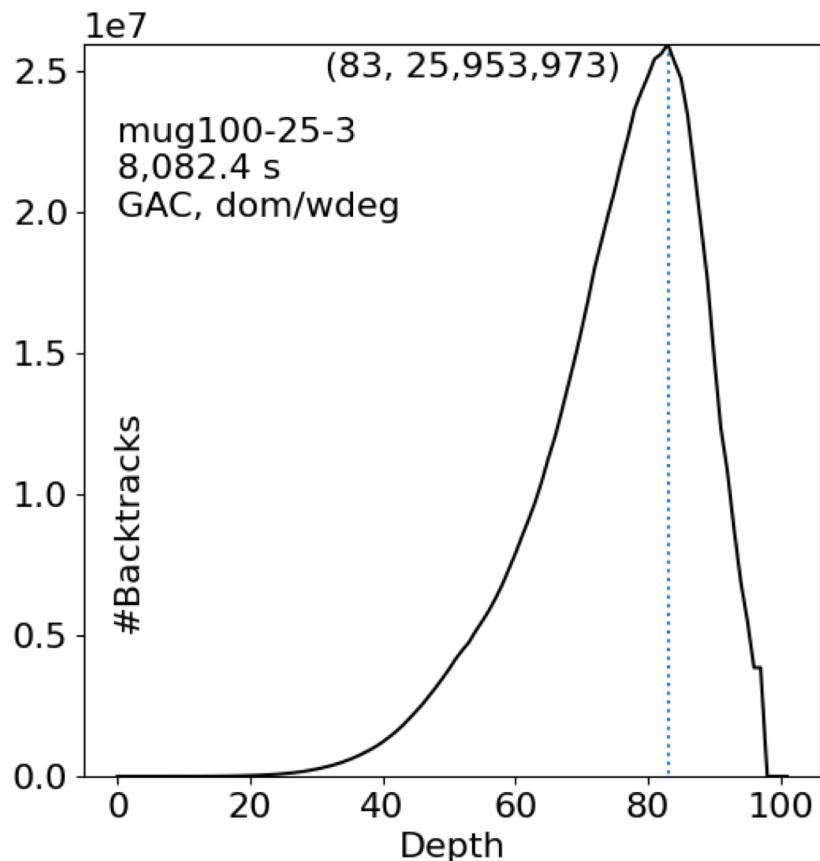
Primal graph of MUG100-25-3

Example: Analyzing structure



Primal graph of MUG100-25-3

Example: Analyzing structure



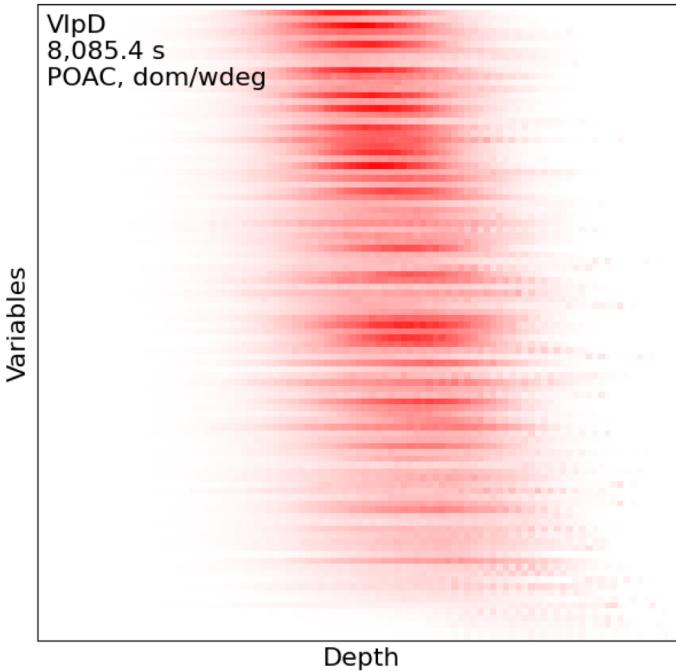
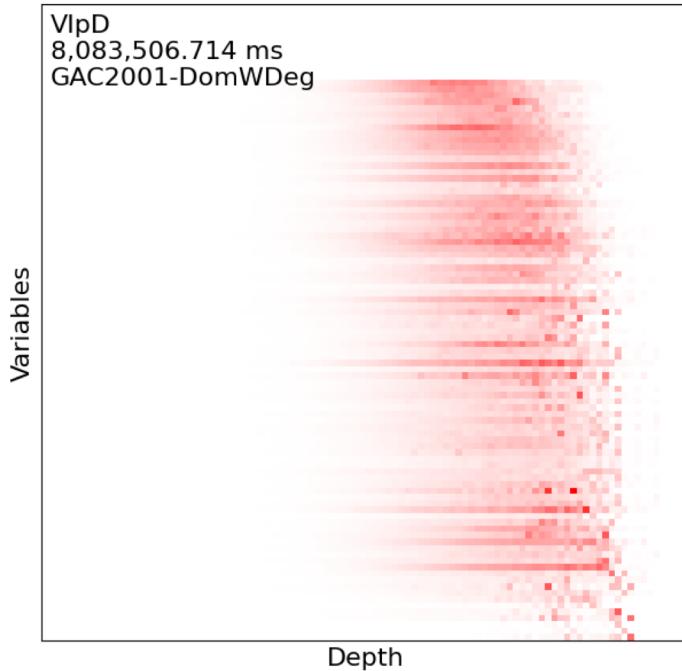
BpD of GAC and DPPC+ on MUG100-25-3

Visualizing variable ordering

- Variable Instantiations per Depth (VIpD)
 - $I(v, d, t)$ is the number of instantiations of variable v at depth d and time t
- Order the variables of the VIpD according to each variable's weighted depth:

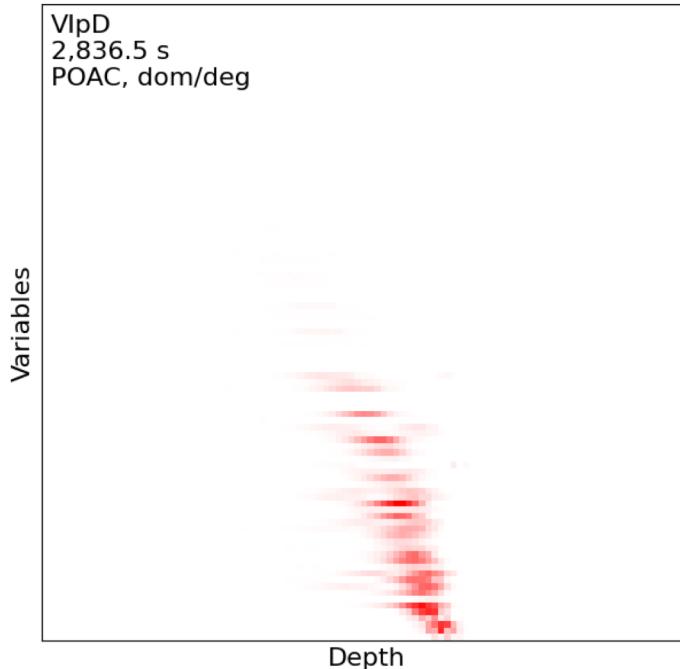
$$d_w(v, t) = \frac{\sum_{d \in D} I(v, d, t) \cdot d}{\sum_{d \in D} I(v, d, t)}$$

Example: Variable ordering



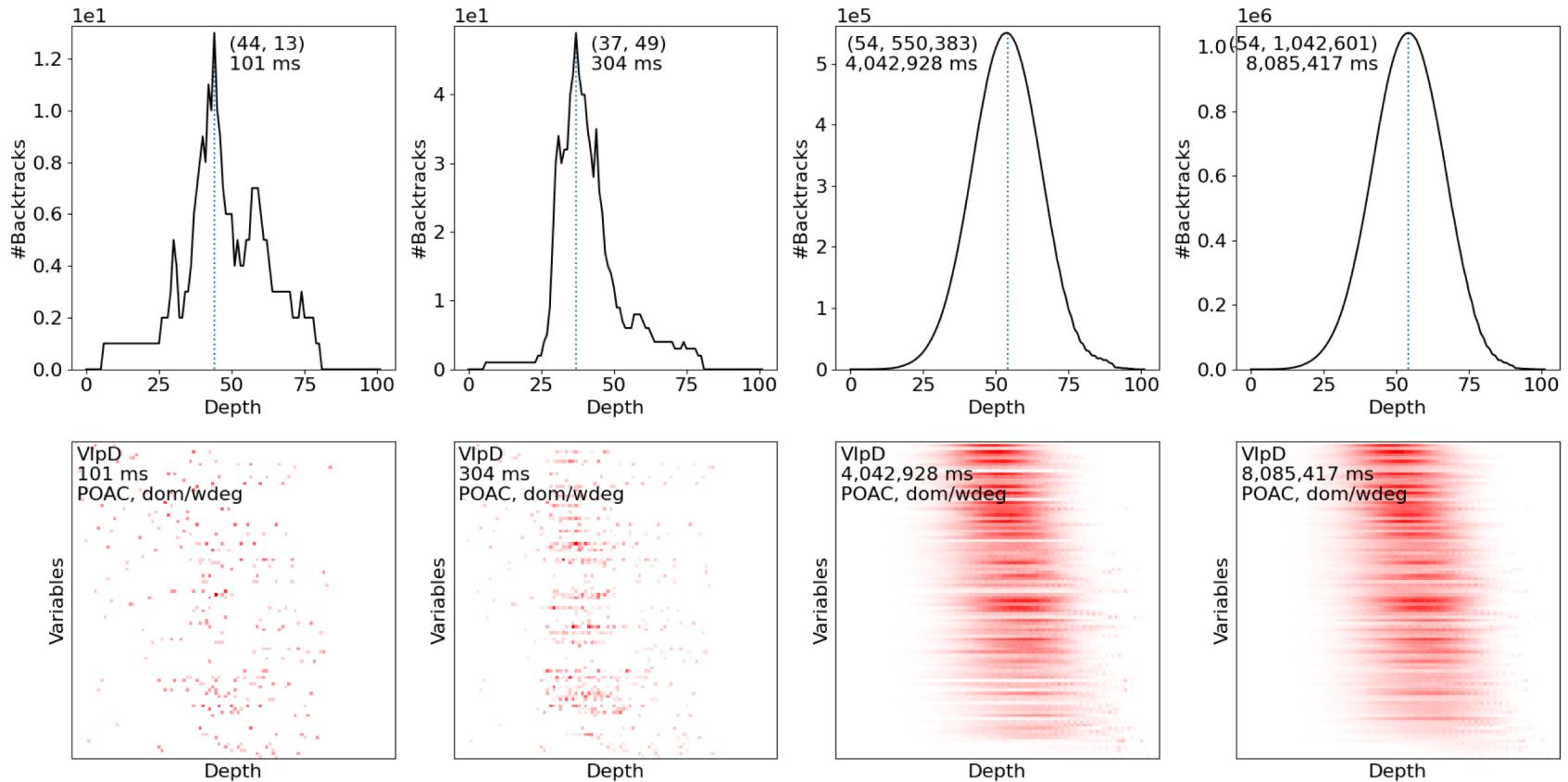
VIpD of GAC and POAC with dom/wdeg on MUG100-1-3

Example: Variable ordering



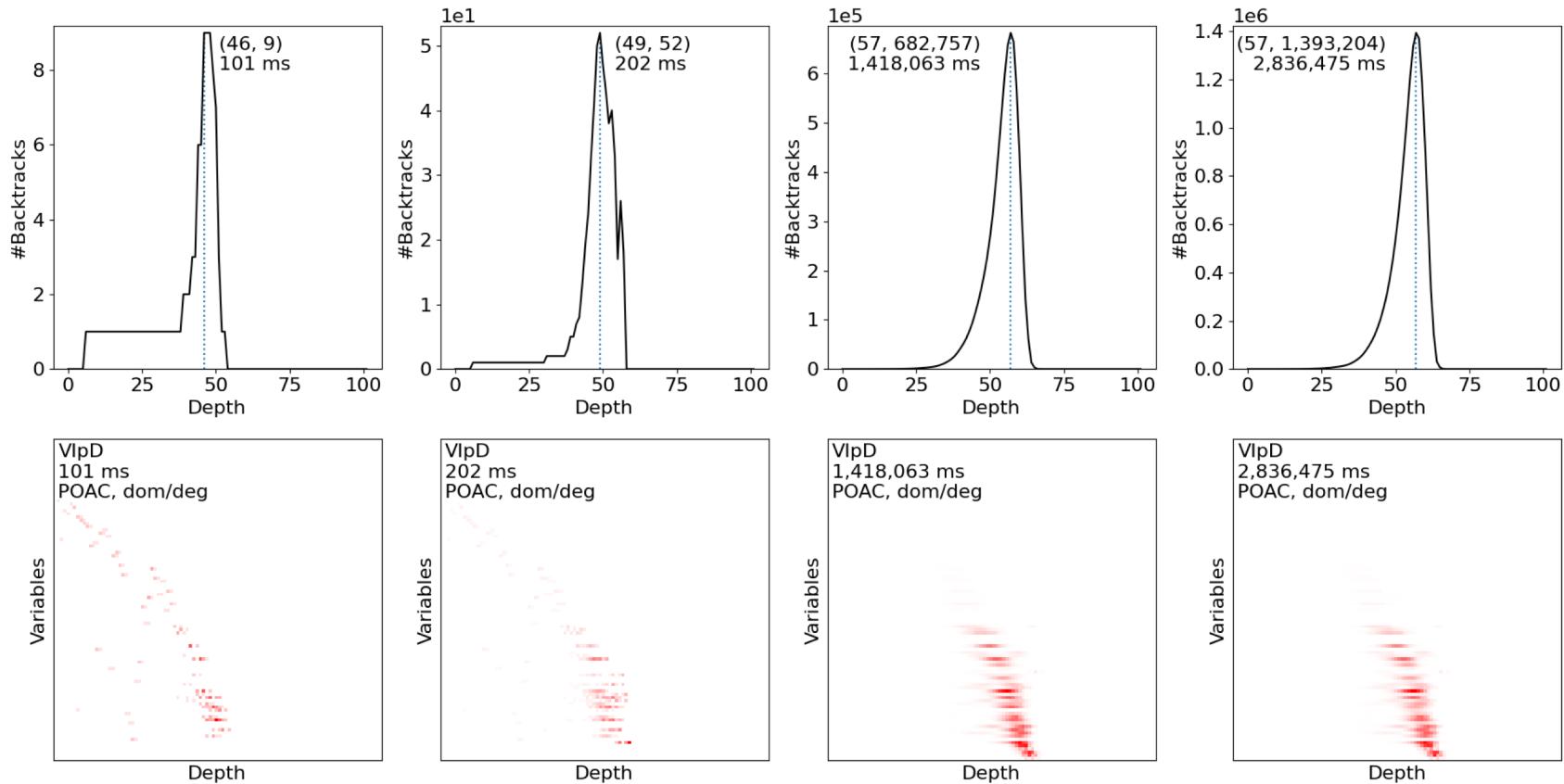
VIpD of POAC with dom/deg on MUG100-1-3

Example: Variable ordering



0.1-summarization of POAC with dom/wdeg on MUG100-1-3

Example: Variable ordering



0.1-summarization of POAC with dom/deg on MUG100-1-3

Conclusions and discussion

- Summarizations can be used to help explain the (changing) behavior of search
- Researchers and developers can use these tools iteratively to better study the impact of a strategy on a given problem
- Summarizations can catch small and large behavior that a human could not

Conclusions and discussion

- Initial results of BpD with binary branching give similar results
- We currently provide a ‘post-mortem’ analysis. Future work includes an ‘in-vivo’ analysis that enables human experts

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