

Foundations of Artificial Intelligence

24. Constraint Satisfaction Problems: Backtracking

Malte Helmert

University of Basel

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Constraint Satisfaction Problems: Overview

Chapter overview: constraint satisfaction problems:

- 22.–23. Introduction
- 24.–26. Basic Algorithms
 - 24. Backtracking
 - 25. Arc Consistency
 - 26. Path Consistency
- 27.–28. Problem Structure

CSP Algorithms

CSP Algorithms

In the following chapters, we consider **algorithms for solving** constraint networks.

basic concepts:

- **search**: check partial assignments systematically
- **backtracking**: discard inconsistent partial assignments
- **inference**: derive equivalent, but tighter constraints to reduce the size of the search space

Naive Backtracking

Naive Backtracking (= Without Inference)

```
function NaiveBacktracking( $\mathcal{C}, \alpha$ ):
```

```
   $\langle V, \text{dom}, (R_{uv}) \rangle := \mathcal{C}$ 
```

```
  if  $\alpha$  is inconsistent with  $\mathcal{C}$ :  
    return inconsistent
```

```
  if  $\alpha$  is a total assignment:  
    return  $\alpha$ 
```

```
  select some variable  $v$  for which  $\alpha$  is not defined
```

```
  for each  $d \in \text{dom}(v)$  in some order:
```

```
     $\alpha' := \alpha \cup \{v \mapsto d\}$ 
```

```
     $\alpha'' := \text{NaiveBacktracking}(\mathcal{C}, \alpha')$ 
```

```
    if  $\alpha'' \neq \text{inconsistent}$ :  
      return  $\alpha''$ 
```

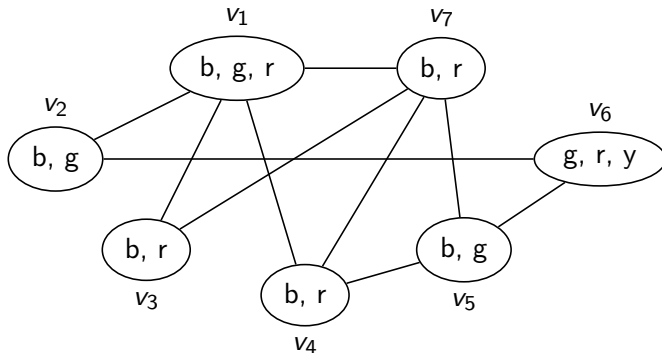
```
  return inconsistent
```

input: constraint network $\mathcal{C}$ and partial assignment α for $\mathcal{C}$
(first invocation: empty assignment $\alpha = \emptyset$)

result: solution of $\mathcal{C}$ or **inconsistent**

Naive Backtracking: Example

Consider the constraint network for the following graph coloring instance:



Naive Backtracking: Example

search tree for naive backtracking with

- **fixed variable order** $v_1, v_7, v_4, v_5, v_6, v_3, v_2$
- **alphabetical** order of the values

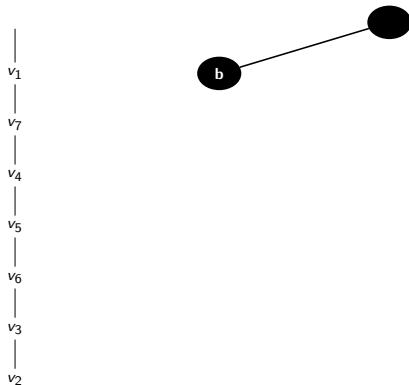
|
 v_1
|
 v_7
|
 v_4
|
 v_5
|
 v_6
|
 v_3
|
 v_2



Naive Backtracking: Example

search tree for naive backtracking with

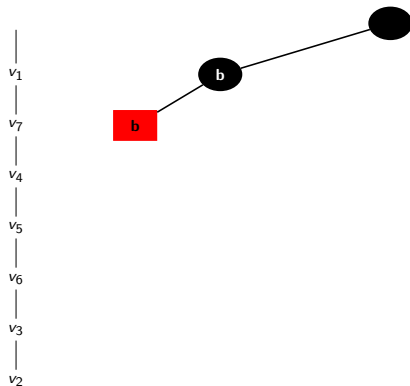
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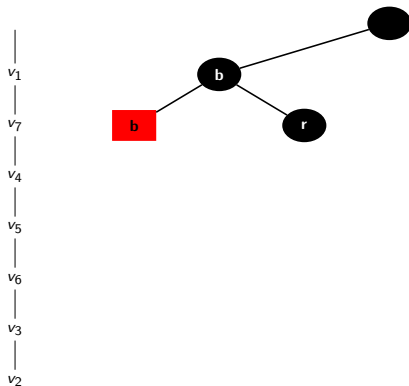
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Naive Backtracking: Example

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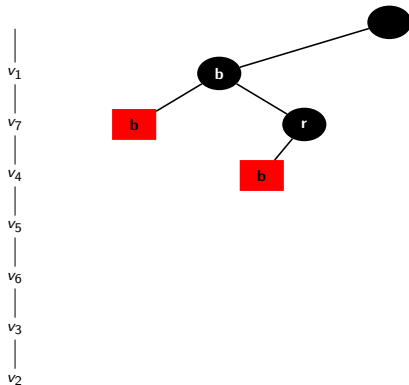
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Naive Backtracking: Example

search tree for naive backtracking with

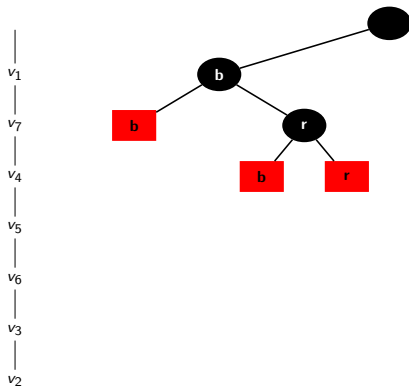
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Naive Backtracking: Example

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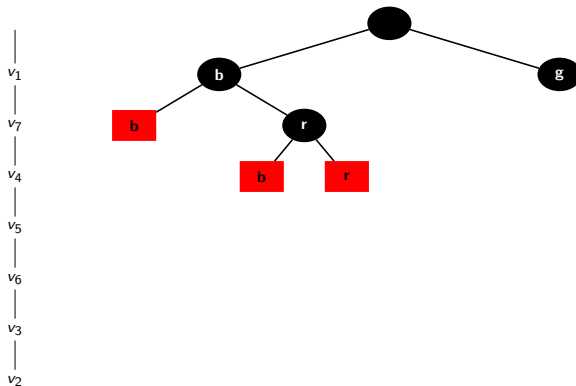
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Naive Backtracking: Example

search tree for naive backtracking with

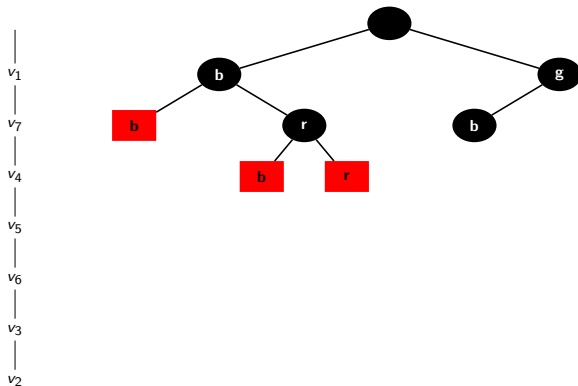
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Naive Backtracking: Example

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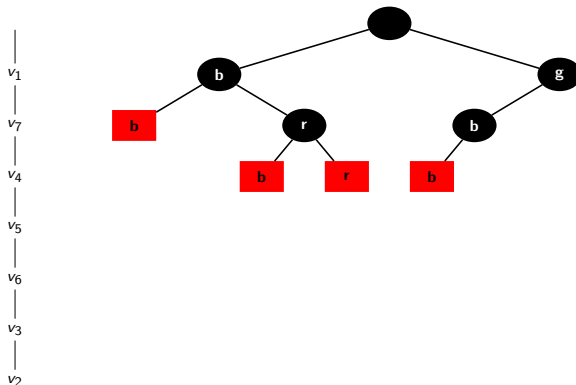
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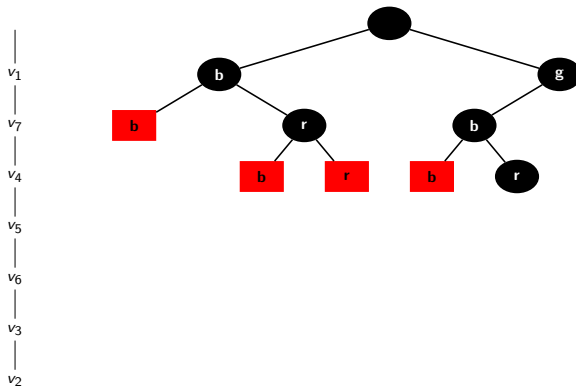
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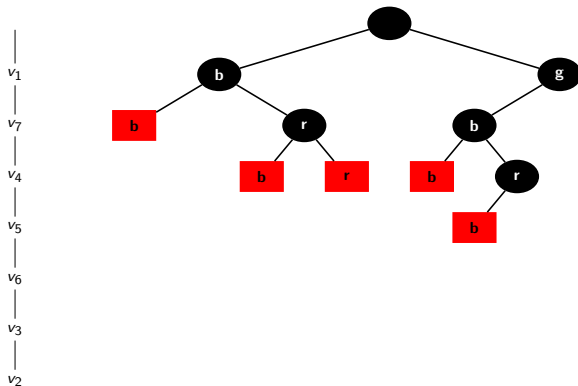
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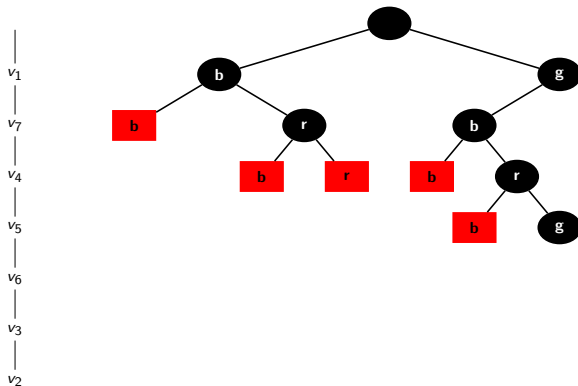
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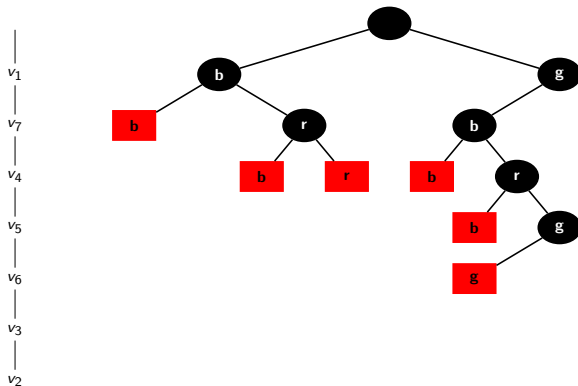
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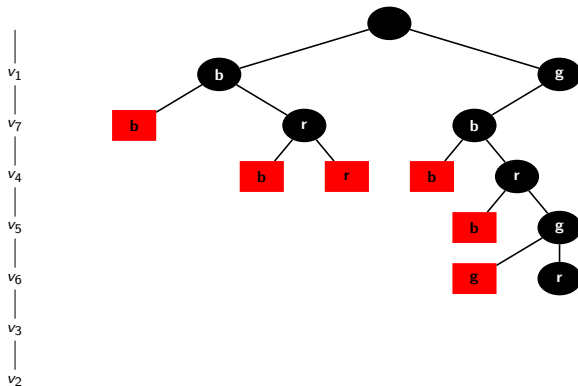
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Naive Backtracking: Example

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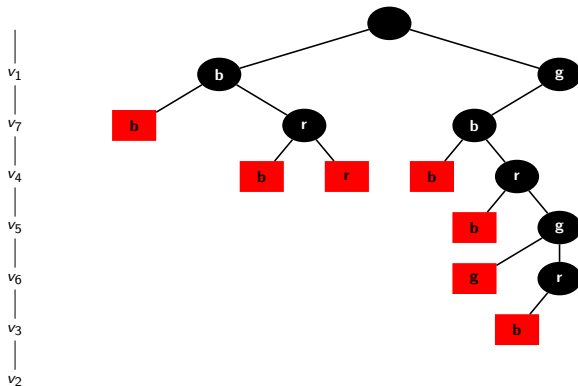
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Naive Backtracking: Example

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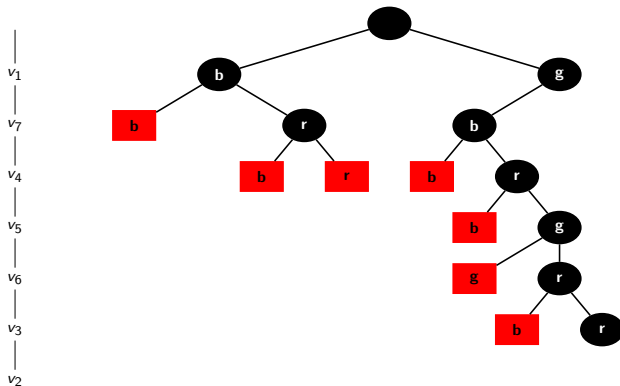
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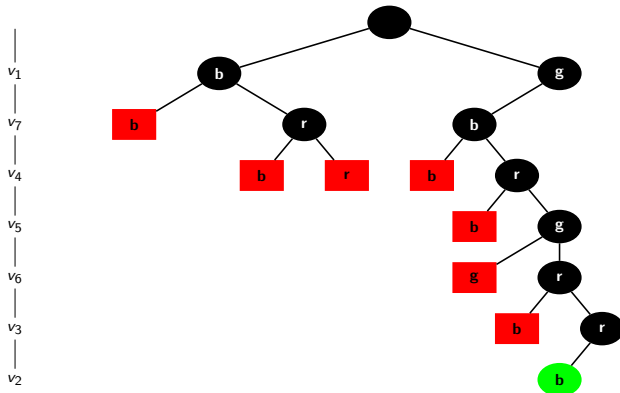


Naive Backtracking: Example

search tree for naive backtracking with

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Is This a New Algorithm?

We have already seen this algorithm:

Backtracking corresponds to depth-first search (Chapter 12)
with the following state space:

- **states**: partial assignments
- **initial state**: empty assignment $\emptyset$
- **goal states**: consistent total assignments
- **actions**: $assign_{v,d}$ assigns value $d \in \text{dom}(v)$ to variable v
- **action costs**: all 0 (all solutions are of equal quality)
- **transitions**:
 - for each **non-total consistent** assignment α ,
choose variable $v = \text{select}(\alpha)$ that is unassigned in α
 - transition $\alpha \xrightarrow{assign_{v,d}} \alpha \cup \{v \mapsto d\}$ for each $d \in \text{dom}(v)$

Why Depth-First Search?

Depth-first search is particularly well-suited for CSPs:

- path length **bounded** (by the number of variables)
- solutions located at **the same depth** (lowest search layer)
- state space is directed **tree**, initial state is the root
 $\rightsquigarrow$ **no duplicates** (Why?)

Hence none of the problematic cases for depth-first search occurs.

Naive Backtracking: Discussion

- Naive backtracking often has to exhaustively explore **similar** search paths (i.e., partial assignments that are identical except for a few variables).
 - “Critical” variables are not recognized and hence considered for assignment (too) late.
 - Decisions that necessarily lead to constraint violations are only recognized when all variables involved in the constraint have been assigned.
- ⇒ more intelligence by **focusing on critical decisions** and by **inference** of consequences of previous decisions

Variable and Value Orders

Naive Backtracking

function NaiveBacktracking($\mathcal{C}, \alpha$):

$\langle V, \text{dom}, (R_{uv}) \rangle := \mathcal{C}$

if α is inconsistent with $\mathcal{C}$:

return inconsistent

if α is a total assignment:

return α

select **some variable** v for which α is not defined

for each $d \in \text{dom}(v)$ **in some order**:

$\alpha' := \alpha \cup \{v \mapsto d\}$

$\alpha'' := \text{NaiveBacktracking}(\mathcal{C}, \alpha')$

if $\alpha'' \neq \text{inconsistent}$:

return α''

return inconsistent

Variable Orders

- Backtracking does not specify in which order **variables** are considered for assignment.
- Such orders can strongly influence the search space size and hence the search performance.
 ↪ **example:** exercises

German: Variablenordnung

Value Orders

- Backtracking does not specify in which order the **values** of the selected variable v are considered.
- This is not as important because it **does not matter** in subtrees without a solution. (**Why not?**)
- **If** there is a solution in the subtree, then ideally a value that leads to a solution should be chosen. (**Why?**)

German: Werteordnung

Static vs. Dynamic Orders

we distinguish:

- **static** orders (fixed prior to search)
- **dynamic** orders (selected variable or value order depends on the search state)

comparison:

- dynamic orders obviously more powerful
- static orders $\rightsquigarrow$ no computational overhead during search

The following ordering criteria can be used statically, but are more effective combined with inference ($\rightsquigarrow$ later) and used dynamically.

Variable Orders

two common variable ordering criteria:

- **minimum remaining values:**
prefer variables that have small **domains**
 - **intuition:** few subtrees $\rightsquigarrow$ smaller tree
 - **extreme case:** only **one** value $\rightsquigarrow$ forced assignment
- **most constraining variable:**
prefer variables contained in **many** nontrivial constraints
 - **intuition:** constraints tested early
 $\rightsquigarrow$ inconsistencies recognized early $\rightsquigarrow$ smaller tree

combination: use minimum remaining values criterion,
then most constraining variable criterion to break ties

Value Orders

Definition (conflict)

Let $\mathcal{C} = \langle V, \text{dom}, (R_{uv}) \rangle$ be a constraint network.

For variables $v \neq v'$ and values $d \in \text{dom}(v)$, $d' \in \text{dom}(v')$, the assignment $v \mapsto d$ is **in conflict** with $v' \mapsto d'$ if $\langle d, d' \rangle \notin R_{vv'}$.

value ordering criterion for partial assignment α
and selected variable v :

- **minimum conflicts:** prefer values $d \in \text{dom}(v)$ such that $v \mapsto d$ causes as few conflicts as possible with variables that are unassigned in α

Summary

Summary: Backtracking

basic search algorithm for constraint networks: **backtracking**

- extends the (initially empty) partial assignment step by step until an **inconsistency** or a **solution** is found
- is a form of **depth-first search**
- depth-first search particularly well-suited because state space is directed tree and all solutions at same (known) depth

Summary: Variable and Value Orders

- **Variable orders** influence the performance of backtracking significantly.
 - goal: **critical** decisions as early as possible
- **Value orders** influence the performance of backtracking on **solvable** constraint networks significantly.
 - goal: **most promising** assignments first