

## Constraint Satisfaction Problems (CSP)

(Where we postpone making difficult decisions until they become easy to make)

R&N: Chap. 5

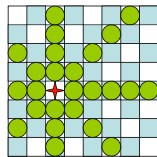
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## What we will try to do ...

- Search techniques make choices in an often arbitrary order. Often little information is available to make each of them
- In many problems, the same states can be reached independent of the order in which choices are made ("commutative" actions)
- Can we solve such problems more efficiently by picking the order appropriately? Can we even avoid making any choice?

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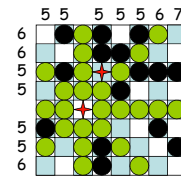
## Constraint Propagation



- Place a queen in a square
- Remove the attacked squares from future consideration

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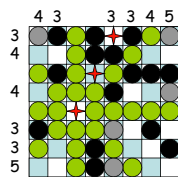
## Constraint Propagation



- Count the number of non-attacked squares in every row and column
- Place a queen in a row or column with minimum number
- Remove the attacked squares from future consideration

4

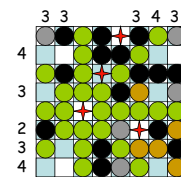
## Constraint Propagation



- Repeat

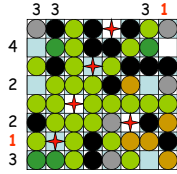
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## Constraint Propagation



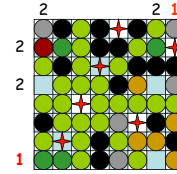
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## Constraint Propagation



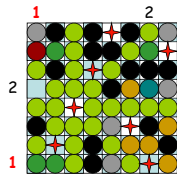
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## Constraint Propagation



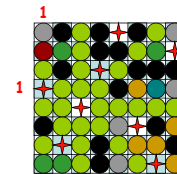
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## Constraint Propagation



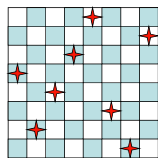
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## Constraint Propagation



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## Constraint Propagation



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## What do we need?

- More than just a successor function and a goal test
- We also need:
  - A means to propagate the constraints imposed by one queen's position on the positions of the other queens
  - An early failure test

→ Explicit representation of constraints  
 → Constraint propagation algorithms

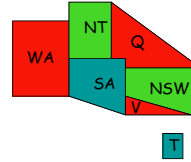
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## Constraint Satisfaction Problem (CSP)

- Set of **variables**  $\{X_1, X_2, \dots, X_n\}$
- Each variable  $X_i$  has a **domain**  $D_i$  of possible values. Usually,  $D_i$  is finite
- Set of **constraints**  $\{C_1, C_2, \dots, C_p\}$
- Each constraint relates a subset of variables by specifying the valid combinations of their values
- Goal: Assign a value to every variable such that all constraints are satisfied

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## Map Coloring



- 7 variables  $\{WA, NT, SA, Q, NSW, V, T\}$
- Each variable has the same domain:  $\{\text{red, green, blue}\}$
- No two adjacent variables have the same value:  
 $WA \neq NT, WA \neq SA, NT \neq SA, NT \neq Q, SA \neq Q,$   
 $SA \neq NSW, SA \neq V, Q \neq NSW, NSW \neq V$

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## 8-Queen Problem

- 8 variables  $X_i, i = 1$  to 8
- The domain of each variable is:  $\{1, 2, \dots, 8\}$
- Constraints are of the forms:
  - $X_i = k \Rightarrow X_j \neq k$  for all  $j = 1$  to 8,  $j \neq i$
  - Similar constraints for diagonals

All constraints are binary

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## Street Puzzle

1 2 3 4 5

$N_i = \{\text{English, Spaniard, Japanese, Italian, Norwegian}\}$   
 $C_i = \{\text{Red, Green, White, Yellow, Blue}\}$   
 $D_i = \{\text{Tea, Coffee, Milk, Fruit-juice, Water}\}$   
 $J_i = \{\text{Painter, Sculptor, Diplomat, Violinist, Doctor}\}$   
 $A_i = \{\text{Dog, Snails, Fox, Horse, Zebra}\}$

The Englishman lives in the Red house  
 The Spaniard has a Dog  
 The Japanese is a Painter  
 The Italian drinks Tea  
 The Norwegian lives in the first house on the left  
 The owner of the Green house drinks Coffee  
 The Green house is on the right of the White house  
 The Sculptor breeds Snails  
 The Diplomat lives in the Yellow house  
 The owner of the middle house drinks Milk  
 The Norwegian lives next door to the Blue house  
 The Violinist drinks Fruit juice  
 The Fox is in the house next to the Doctor's  
 The Horse is next to the Diplomat's

Who owns the Zebra?  
 Who drinks Water?

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 The Fox is in the house next to the Doctor's  
 The Horse is next to the Diplomat's

$\forall i, j \in [1, 5], i \neq j, N_i \neq N_j$   
 $\forall i, j \in [1, 5], i \neq j, C_i \neq C_j$   
 ...

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## Street Puzzle

1 2 3 4 5

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The Englishman lives in the Red house  $\dots \rightarrow (N_i = \text{English}) \Leftrightarrow (C_i = \text{Red})$   
 The Spaniard has a Dog  
 The Japanese is a Painter  $\dots \rightarrow (N_i = \text{Japanese}) \Leftrightarrow (J_i = \text{Painter})$   
 The Italian drinks Tea  
 The Norwegian lives in the first house on the left  $\dots \rightarrow (N_i = \text{Norwegian})$   
 The owner of the Green house drinks Coffee  
 The Green house is on the right of the White house  
 The Sculptor breeds Snails  
 The Diplomat lives in the Yellow house  
 The owner of the middle house drinks Milk  
 The Norwegian lives next door to the Blue house  
 The Violinist drinks Fruit juice  
 The Fox is in the house next to the Doctor's  
 The Horse is next to the Diplomat's

$(C_i = \text{White}) \Leftrightarrow (C_{i+1} = \text{Green})$   
 $(C_5 \neq \text{White})$   
 $(C_1 \neq \text{Green})$

left as an exercise 18

## Street Puzzle

1 2 3 4 5

$N_i = \{\text{English, Spaniard, Japanese, Italian, Norwegian}\}$   
 $C_i = \{\text{Red, Green, White, Yellow, Blue}\}$   
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The Englishman lives in the Red house  $\rightarrow (N_i = \text{English}) \Leftrightarrow (C_i = \text{Red})$   
 The Spaniard has a Dog  
 The Japanese is a Painter  $\rightarrow (N_i = \text{Japanese}) \Leftrightarrow (J_i = \text{Painter})$   
 The Italian drinks Tea  
 The Norwegian lives in the first house on the left  $\rightarrow (N_i = \text{Norwegian})$   
 The owner of the Green house drinks Coffee  
 The Green house is on the right of the White house  
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$(C_i = \text{White}) \Leftrightarrow (C_{i+1} = \text{Green})$   
 $(C_5 \neq \text{White})$   
 $(C_1 \neq \text{Green})$

Unary constraints

## Street Puzzle

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The Englishman lives in the Red house  
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 The Japanese is a Painter  
 The Italian drinks Tea  
 The Norwegian lives in the first house on the left  $\rightarrow N_i = \text{Norwegian}$   
 The owner of the Green house drinks Coffee  
 The Green house is on the right of the White house  
 The Sculptor breeds Snails  
 The Diplomat lives in the Yellow house  
 The owner of the middle house drinks Milk  $\rightarrow D_3 = \text{Milk}$   
 The Norwegian lives next door to the Blue house  
 The Violinist drinks Fruit juice  
 The Fox is in the house next to the Doctor's  
 The Horse is next to the Diplomat's

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## Street Puzzle

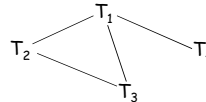
1 2 3 4 5

$N_i = \{\text{English, Spaniard, Japanese, Italian, Norwegian}\}$   
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The Englishman lives in the Red house  $\rightarrow C_i \neq \text{Red}$   
 The Spaniard has a Dog  $\rightarrow A_i \neq \text{Dog}$   
 The Japanese is a Painter  
 The Italian drinks Tea  
 The Norwegian lives in the first house on the left  $\rightarrow N_i = \text{Norwegian}$   
 The owner of the Green house drinks Coffee  
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 The Diplomat lives in the Yellow house  
 The owner of the middle house drinks Milk  $\rightarrow D_3 = \text{Milk}$   
 The Norwegian lives next door to the Blue house  
 The Violinist drinks Fruit juice  $\rightarrow J_3 \neq \text{Violinist}$   
 The Fox is in the house next to the Doctor's  
 The Horse is next to the Diplomat's

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## Task Scheduling



Four tasks  $T_1, T_2, T_3,$  and  $T_4$  are related by time constraints:

- $T_1$  must be done during  $T_3$
- $T_2$  must be achieved before  $T_1$  starts
- $T_2$  must overlap with  $T_3$
- $T_4$  must start after  $T_1$  is complete
- Are the constraints compatible?
- What are the possible time relations between two tasks?
- What if the tasks use resources in limited supply?

How to formulate this problem as a CSP?

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## 3-SAT

- $n$  Boolean variables  $u_1, \dots, u_n$
- $p$  constraints of the form  $u_i^* \vee u_j^* \vee u_k^* = 1$  where  $u^*$  stands for either  $u$  or  $\neg u$
- Known to be NP-complete

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## Finite vs. Infinite CSP

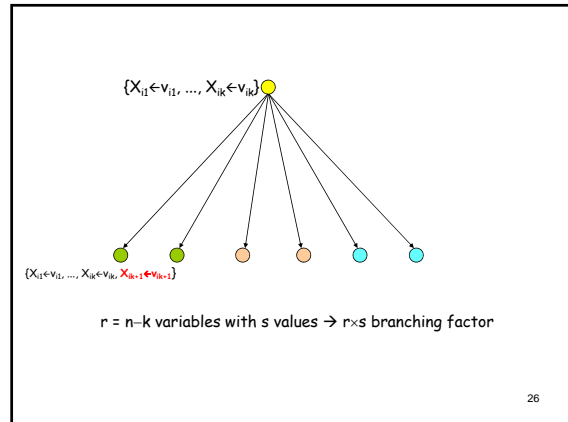
- Finite CSP:** each variable has a finite domain of values
- Infinite CSP:** some or all variables have an infinite domain  
 E.g., linear programming problems over the reals:  
 for  $i = 1, 2, \dots, p$ :  $a_{i,1}x_1 + a_{i,2}x_2 + \dots + a_{i,n}x_n = a_{i,0}$   
 for  $j = 1, 2, \dots, q$ :  $b_{j,1}x_1 + b_{j,2}x_2 + \dots + b_{j,n}x_n \leq b_{j,0}$
- We will only consider finite CSP

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## CSP as a Search Problem

- $n$  variables  $X_1, \dots, X_n$
- **Valid assignment:**  $\{X_{i1} \leftarrow v_{i1}, \dots, X_{ik} \leftarrow v_{ik}\}$ ,  $0 \leq k \leq n$ , such that the values  $v_{i1}, \dots, v_{ik}$  satisfy all constraints relating the variables  $X_{i1}, \dots, X_{ik}$
- **Complete assignment:** one where  $k = n$   
[if all variable domains have size  $d$ , there are  $O(d^n)$  complete assignments]
- **States:** valid assignments
- **Initial state:** empty assignment  $\{\}$ , i.e.  $k = 0$
- **Successor of a state:**  
 $\{X_{i1} \leftarrow v_{i1}, \dots, X_{ik} \leftarrow v_{ik}\} \rightarrow \{X_{i1} \leftarrow v_{i1}, \dots, X_{ik} \leftarrow v_{ik}, X_{i(k+1)} \leftarrow v_{i(k+1)}\}$
- **Goal test:**  $k = n$

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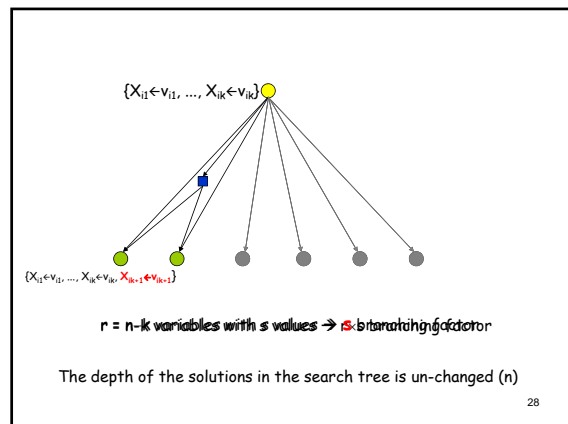
## A Key property of CSP: Commutativity

The order in which variables are assigned values has no impact on the reachable complete valid assignments

Hence:

- 1) One can expand a node  $N$  by first selecting **one** variable  $X$  not in the assignment  $A$  associated with  $N$  and then assigning every value  $v$  in the domain of  $X$   
[ $\rightarrow$  big reduction in branching factor]

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- 4 variables  $X_1, \dots, X_4$
- Let the valid assignment of  $N$  be:  
 $A = \{X_1 \leftarrow v_1, X_3 \leftarrow v_3\}$
- For example pick variable  $X_4$
- Let the domain of  $X_4$  be  $\{v_{4,1}, v_{4,2}, v_{4,3}\}$
- The successors of  $A$  are all the valid assignments among:
  - $\{X_1 \leftarrow v_1, X_3 \leftarrow v_3, X_4 \leftarrow v_{4,1}\}$
  - $\{X_1 \leftarrow v_1, X_3 \leftarrow v_3, X_4 \leftarrow v_{4,2}\}$
  - $\{X_1 \leftarrow v_1, X_3 \leftarrow v_3, X_4 \leftarrow v_{4,3}\}$

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## A Key property of CSP: Commutativity

The order in which variables are assigned values has no impact on the reachable complete valid assignments

Hence:

- 1) One can expand a node  $N$  by first selecting **one** variable  $X$  not in the assignment  $A$  associated with  $N$  and then assigning every value  $v$  in the domain of  $X$   
[ $\rightarrow$  big reduction in branching factor]
- 2) One need not store the path to a node  
 $\rightarrow$  **Backtracking** search algorithm

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## Backtracking Search

Essentially a simplified depth-first algorithm using recursion

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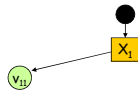
## Backtracking Search (3 variables)



Assignment = {}

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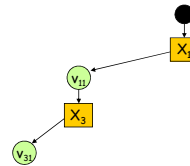
## Backtracking Search (3 variables)



Assignment = {(X<sub>1</sub>, v<sub>11</sub>)}

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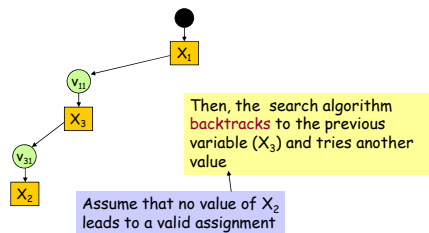
## Backtracking Search (3 variables)



Assignment = {(X<sub>1</sub>, v<sub>11</sub>), (X<sub>3</sub>, v<sub>31</sub>)}

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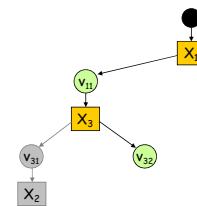
## Backtracking Search (3 variables)



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## Backtracking Search (3 variables)



Assignment = {(X<sub>1</sub>, v<sub>11</sub>), (X<sub>3</sub>, v<sub>32</sub>)}

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### Backtracking Search (3 variables)

The search algorithm backtracks to the previous variable ( $X_3$ ) and tries another value. But assume that  $X_3$  has only two possible values. The algorithm backtracks to  $X_1$ .

Assume again that no value of  $X_2$  leads to a valid assignment

Assignment =  $\{(X_1, v_{11}), (X_3, v_{32})\}$

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### Backtracking Search (3 variables)

Assignment =  $\{(X_1, v_{12})\}$

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### Backtracking Search (3 variables)

Assignment =  $\{(X_1, v_{12}), (X_2, v_{21})\}$

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### Backtracking Search (3 variables)

The algorithm need not consider the variables in the same order in this sub-tree as in the other

Assignment =  $\{(X_1, v_{12}), (X_2, v_{21})\}$

40

### Backtracking Search (3 variables)

Assignment =  $\{(X_1, v_{12}), (X_2, v_{21}), (X_3, v_{32})\}$

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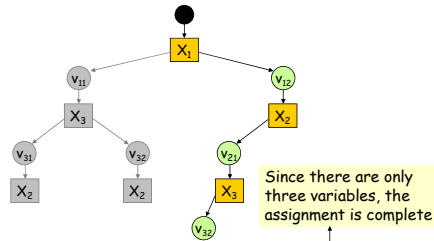
### Backtracking Search (3 variables)

The algorithm need not consider the values of  $X_3$  in the same order in this sub-tree

Assignment =  $\{(X_1, v_{12}), (X_2, v_{21}), (X_3, v_{32})\}$

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### Backtracking Search (3 variables)



Assignment =  $\{(X_1, v_{12}), (X_2, v_{21}), (X_3, v_{32})\}$

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### Backtracking Algorithm

$CSP\text{-}BACKTRACKING(A)$

1. If assignment  $A$  is complete then return  $A$
2.  $X \leftarrow$  select a variable not in  $A$
3.  $D \leftarrow$  select an ordering on the domain of  $X$
4. For each value  $v$  in  $D$  do
  - a. Add  $(X \leftarrow v)$  to  $A$
  - b. If  $A$  is valid then
    - i.  $result \leftarrow CSP\text{-}BACKTRACKING(A)$
    - ii. If  $result \neq failure$  then return  $result$
  - c. Remove  $(X \leftarrow v)$  from  $A$
5. Return failure

Call  $CSP\text{-}BACKTRACKING(\{\})$

[This recursive algorithm keeps too much data in memory. An iterative version could save memory (left as an exercise)]

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### Critical Questions for the Efficiency of CSP-Backtracking

$CSP\text{-}BACKTRACKING(A)$

1. If assignment  $A$  is complete then return  $A$
2.  $X \leftarrow$  **select** a variable not in  $A$
3.  $D \leftarrow$  **select** an ordering on the domain of  $X$
4. For each value  $v$  in  $D$  do
  - a. Add  $(X \leftarrow v)$  to  $A$
  - b. If  $a$  is valid then
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    - ii. If  $result \neq failure$  then return  $result$
  - c. Remove  $(X \leftarrow v)$  from  $A$
5. Return failure

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### Critical Questions for the Efficiency of CSP-Backtracking

1) Which variable  $X$  should be assigned a value next?

2) In which order should  $X$ 's values be assigned?

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### Critical Questions for the Efficiency of CSP-Backtracking

- 1) Which variable  $X$  should be assigned a value next?  
The current assignment may not lead to any solution, but the algorithm does not know it yet. Selecting the right variable  $X$  may help discover the contradiction more quickly
- 2) In which order should  $X$ 's values be assigned?

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### Critical Questions for the Efficiency of CSP-Backtracking

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- 2) In which order should  $X$ 's values be assigned?  
The current assignment may be part of a solution. Selecting the right value to assign to  $X$  may help discover this solution more quickly

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### Critical Questions for the Efficiency of CSP-Backtracking

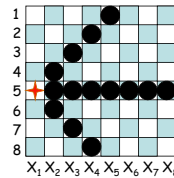
- 1) Which variable  $X$  should be assigned a value next?  
The current assignment may not lead to any solution, but the algorithm does not know it yet. Selecting the right variable  $X$  may help discover the contradiction more quickly
- 2) In which order should  $X$ 's values be assigned?  
The current assignment may be part of a solution. Selecting the right value to assign to  $X$  may help discover this solution more quickly

More on these questions very soon ...

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### Forward Checking

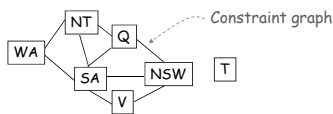
A simple constraint-propagation technique:



Assigning the value 5 to  $X_1$  leads to removing values from the domains of  $X_2, X_3, \dots, X_8$

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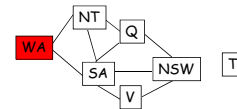
### Forward Checking in Map Coloring



WA	NT	Q	NSW	V	SA	T
RGB	RGB	RGB	RGB	RGB	RGB	RGB

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### Forward Checking in Map Coloring

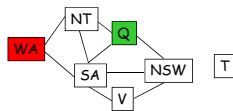


WA	NT	Q	NSW	V	SA	T
RGB	RGB	RGB	RGB	RGB	RGB	RGB
R	<del>RGB</del>	RGB	RGB	RGB	<del>RGB</del>	RGB

Forward checking removes the value Red of NT and of SA

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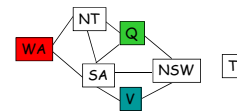
### Forward Checking in Map Coloring



WA	NT	Q	NSW	V	SA	T
RGB	RGB	RGB	RGB	RGB	RGB	RGB
R	GB	RGB	RGB	RGB	GB	RGB
R	<del>B</del>	G	<del>B</del>	RGB	<del>B</del>	RGB

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### Forward Checking in Map Coloring



WA	NT	Q	NSW	V	SA	T
RGB	RGB	RGB	RGB	RGB	RGB	RGB
R	GB	RGB	RGB	RGB	GB	RGB
R	B	G	RB	RGB	B	RGB
R	B	G	<del>R</del>	B	<del>B</del>	RGB

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## Forward Checking in Map Coloring

Empty set: the current assignment  
 $\{(WA \leftarrow R), (Q \leftarrow G), (V \leftarrow B)\}$   
 does not lead to a solution

WA	NT	Q	NSW	V	SA	T
RGB	RGB	RGB	RGB	RGB	RGB	RGB
R	GB	RGB	RGB	RGB	GB	RGB
R	B	G	RB	RGB	B	RGB
R	B	G	<del>R</del>	B	<del>B</del>	RGB

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## Forward Checking (General Form)

Whenever a pair  $(X \leftarrow v)$  is added to assignment  $A$  do:

For each variable  $Y$  not in  $A$  do:

For every constraint  $C$  relating  $Y$  to  
 the variables in  $A$  do:

Remove all values from  $Y$ 's domain  
 that do not satisfy  $C$

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## Modified Backtracking Algorithm

CSP-BACKTRACKING( $A$ , var-domains)

1. If assignment  $A$  is complete then return  $A$
2.  $X \leftarrow$  select a variable not in  $A$
3.  $D \leftarrow$  select an ordering on the domain of  $X$
4. For each value  $v$  in  $D$  do
  - a. Add  $(X \leftarrow v)$  to  $A$
  - b. var-domains  $\leftarrow$  forward checking(var-domains,  $X$ ,  $v$ ,  $A$ )
  - c. If no variable has an empty domain then
    - (i) result  $\leftarrow$  CSP-BACKTRACKING( $A$ , var-domains)
    - (ii) If result  $\neq$  failure then return result
  - d. Remove  $(X \leftarrow v)$  from  $A$
5. Return failure

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## Modified Backtracking Algorithm

CSP-BACKTRACKING( $A$ , var-domains)

1. If assignment  $A$  is complete then return  $A$
2.  $X \leftarrow$  select a variable not in  $A$
3.  $D \leftarrow$  select an ordering on the domain of  $X$
4. For each value  $v$  in  $D$  do
  - a. Add  $(X \leftarrow v)$  to  $A$  No need any more to verify that  $A$  is valid
  - b. var-domains  $\leftarrow$  forward checking(var-domains,  $X$ ,  $v$ ,  $A$ )
  - c. If no variable has an empty domain then
    - (i) result  $\leftarrow$  CSP-BACKTRACKING( $A$ , var-domains)
    - (ii) If result  $\neq$  failure then return result
  - d. Remove  $(X \leftarrow v)$  from  $A$
5. Return failure

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## Modified Backtracking Algorithm

CSP-BACKTRACKING( $A$ , var-domains)

1. If assignment  $A$  is complete then return  $A$
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    - (i) result  $\leftarrow$  CSP-BACKTRACKING( $A$ , var-domains)
    - (ii) If result  $\neq$  failure then return result
  - d. Remove  $(X \leftarrow v)$  from  $A$
5. Return failure

Need to pass down the updated variable domains

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## Modified Backtracking Algorithm

CSP-BACKTRACKING( $A$ , var-domains)

1. If assignment  $A$  is complete then return  $A$
2.  $X \leftarrow$  select a variable not in  $A$
3.  $D \leftarrow$  select an ordering on the domain of  $X$
4. For each value  $v$  in  $D$  do
  - a. Add  $(X \leftarrow v)$  to  $A$
  - b. var-domains  $\leftarrow$  forward checking(var-domains,  $X$ ,  $v$ ,  $A$ )
  - c. If no variable has an empty domain then
    - (i) result  $\leftarrow$  CSP-BACKTRACKING( $A$ , var-domains)
    - (ii) If result  $\neq$  failure then return result
  - d. Remove  $(X \leftarrow v)$  from  $A$
5. Return failure

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- 1) Which variable  $X_i$  should be assigned a value next?
  - Most-constrained-variable heuristic
  - Most-constraining-variable heuristic
- 2) In which order should its values be assigned?
  - Least-constraining-value heuristic

These heuristics can be quite confusing

Keep in mind that **all** variables must eventually get a value, while only **one** value from a domain must be assigned to each variable

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### Most-Constrained-Variable Heuristic

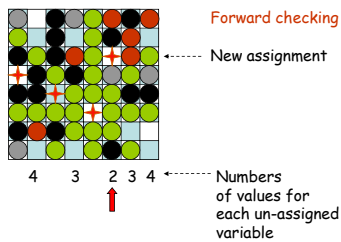
- 1) Which variable  $X_i$  should be assigned a value next?

Select the variable with the smallest remaining domain

[Rationale: Minimize the branching factor]

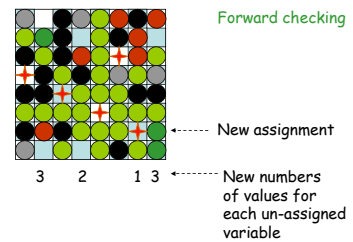
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### 8-Queens



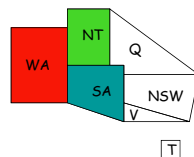
63

### 8-Queens



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### Map Coloring



- SA's remaining domain has size 1 (value Blue remaining)
  - Q's remaining domain has size 2
  - NSW's, V's, and T's remaining domains have size 3
- Select SA

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### Most-Constraining-Variable Heuristic

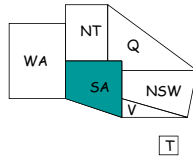
- 1) Which variable  $X_i$  should be assigned a value next?

Among the variables with the smallest remaining domains (ties with respect to the most-constrained-variable heuristic), select the one that appears in the largest number of constraints on variables not in the current assignment

[Rationale: Increase future elimination of values, to reduce future branching factors]

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## Map Coloring



- Before any value has been assigned, all variables have a domain of size 3, but SA is involved in more constraints (5) than any other variable
- Select SA and assign a value to it (e.g., Blue)

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## Least-Constraining-Value Heuristic

2) In which order should X's values be assigned?

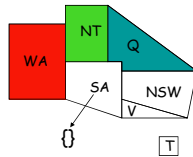
Select the value of X that removes the smallest number of values from the domains of those variables which are not in the current assignment

[Rationale: Since only one value will eventually be assigned to X, pick the least-constraining value first, since it is the most likely not to lead to an invalid assignment]

[Note: Using this heuristic requires performing a forward-checking step for every value, not just for the selected value]

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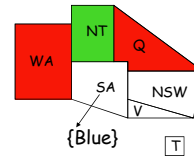
## Map Coloring



- Q's domain has two remaining values: Blue and Red
- Assigning Blue to Q would leave 0 value for SA, while assigning Red would leave 1 value

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## Map Coloring



- Q's domain has two remaining values: Blue and Red
- Assigning Blue to Q would leave 0 value for SA, while assigning Red would leave 1 value
- So, assign Red to Q

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## Modified Backtracking Algorithm

```

CSP-BACKTRACKING(A, var-domains)
1. If assignment A is complete then return A
2. X ← select a variable not in A
3. D ← select an ordering on the domain of X
4. For each value v in D do
    a. Add (X←v) to A
    b. var-domains ← forward checking(var-domains, X, v, A)
    c. If no variable has an empty domain then
       (i) result ← CSP-BACKTRACKING(A, var-domains)
       (ii) If result ≠ failure then return result
    d. Remove (X←v) from A
Return failure
    
```

1) Most-constrained-variable heuristic  
2) Most-constraining-variable heuristic

3) Least-constraining-value heuristic 5.

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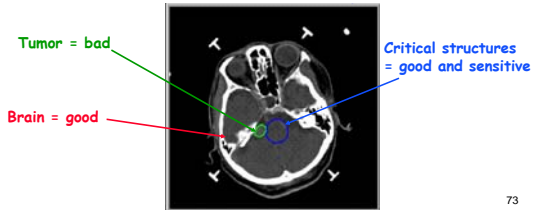
## Applications of CSP

- CSP techniques are widely used
- Applications include:
  - Crew assignments to flights
  - Management of transportation fleet
  - Flight/rail schedules
  - Job shop scheduling
  - Task scheduling in port operations
  - Design, including spatial layout design
  - Radiosurgical procedures

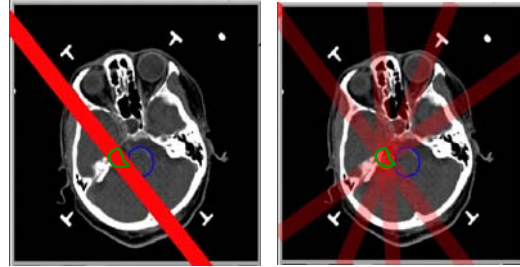
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## Radiosurgery

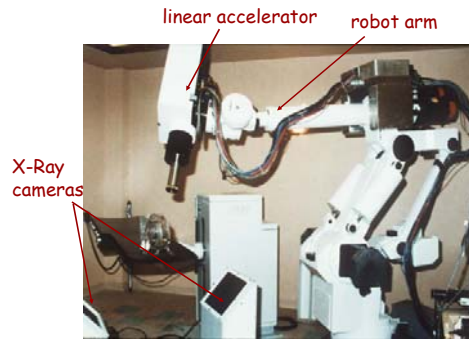
Minimally invasive procedure that uses a beam of radiation as an ablative surgical instrument to destroy tumors



## Problem

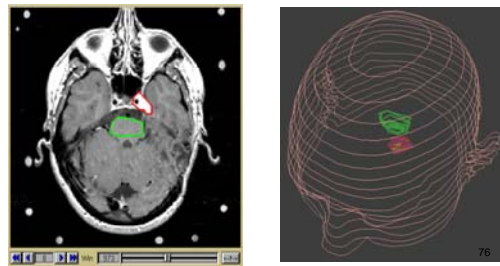


## The CyberKnife



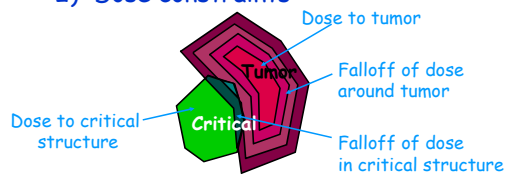
## Inputs

### 1) Regions of interest

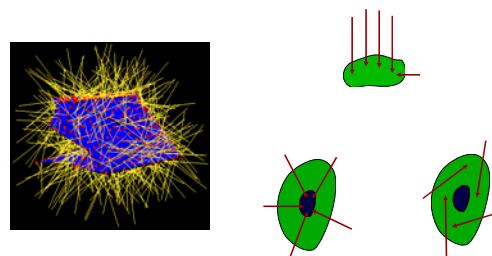


## Inputs

### 2) Dose constraints



## Beam Sampling



### Constraints

- $2000 \leq \text{Tumor} \leq 2200$ 
  - $2000 \leq B2 + B4 \leq 2200$
  - $2000 \leq B4 \leq 2200$
  - $2000 \leq B3 + B4 \leq 2200$
  - $2000 \leq B3 \leq 2200$
  - $2000 \leq B1 + B3 + B4 \leq 2200$
  - $2000 \leq B1 + B4 \leq 2200$
  - $2000 \leq B1 + B2 + B4 \leq 2200$
  - $2000 \leq B1 \leq 2200$
  - $2000 \leq B1 + B2 \leq 2200$
- $0 \leq \text{Critical} \leq 500$ 
  - $0 \leq B2 \leq 500$

$2000 < \text{Tumor} < 2200$

- $2000 < B2 + B4 < 2200$
- $2000 < B4 < 2200$
- $2000 < B3 + B4 < 2200$
- $2000 < B3 < 2200$
- $2000 < B1 + B3 + B4 < 2200$
- $2000 < B1 + B4 < 2200$
- $2000 < B1 + B2 + B4 < 2200$
- $2000 < B1 < 2200$
- $2000 < B1 + B2 < 2200$

$2000 < \text{Tumor} < 2200$

- $2000 < B4$
- $2000 < B3$
- $B1 + B3 + B4 < 2200$
- $B1 + B2 + B4 < 2200$
- $2000 < B1$

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### Case Results

**50% Isodose Surface**

LINAC system

**50% Isodose Surface**

Cyberknife

**80% Isodose Surface**

LINAC system

**80% Isodose Surface**

Cyberknife

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### THE POWER OF T<sup>M</sup> TECHNOLOGY

CyberKnife<sup>TM</sup> System for Linear TV Radiosurgery with UltraSteer<sup>TM</sup> Linac

#### INTEGRATION OF TWO REVOLUTIONARY TECHNOLOGIES

**Proprietary Image-Guidance System**  
 Includes an on-board X-ray imager, a patient immobilization system, and a real-time motion tracking system.

**Multi-Isodose Robotic Arm**  
 Allows you to precisely prescribe dose and deliver it using 6000 tiny beams.

Integration of these two technologies allows physicians to treat complex-shaped tumors with high precision accuracy. The dose is automatically delivered.

**Fast setup, no frameless stereotaxical system.**

#### Simple Outpatient Treatment Process

**Planning:** Using our advanced treatment planning software.

**Positioning:** Patients lie on the table and are immobilized. A high-precision motion tracking system is used to precisely track the patient's position.

**Verification:** The image-guidance system tracks the patient's position in real-time, ensuring that the dose is delivered exactly where it is needed.

**Delivery:** The CyberKnife system delivers the dose to the tumor.

**Recovery:** Patients are able to return to their normal activities of daily living.

**Completion:** The CyberKnife system tracks the patient's position throughout the treatment.

**CyberKnife<sup>TM</sup> Radiosurgery**  
 Treatment of all tumors.

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