

# CSPs: Search and Arc Consistency



Computer Science cpsc322, Lecture 12

*(Textbook Chpt 4.3-4.5)*

January, 30, 2009



# Lecture Overview

- **Recap CSPs**
- Generate-and-Test
- Search 
- Consistency 
- Arc Consistency

# Constraint Satisfaction Problems: definitions

## Definition (Constraint Satisfaction Problem)

A constraint satisfaction problem consists of

- a set of variables

$A, B, C$

$\text{dom } C = \{1, 2, 3\}$

- a domain for each variable

$\text{dom } A = \{1, 2, 3, 4, 5\}$

$\text{dom } B = \text{dom } A$

- a set of constraints

$\rightarrow B = 5$   ~~$C < B$~~

$C < B$

$A = B$

no solutions

# possible worlds  
75

## Definition (model / solution)

A **model** of a CSP is an assignment of values to variables that satisfies all of the constraints.

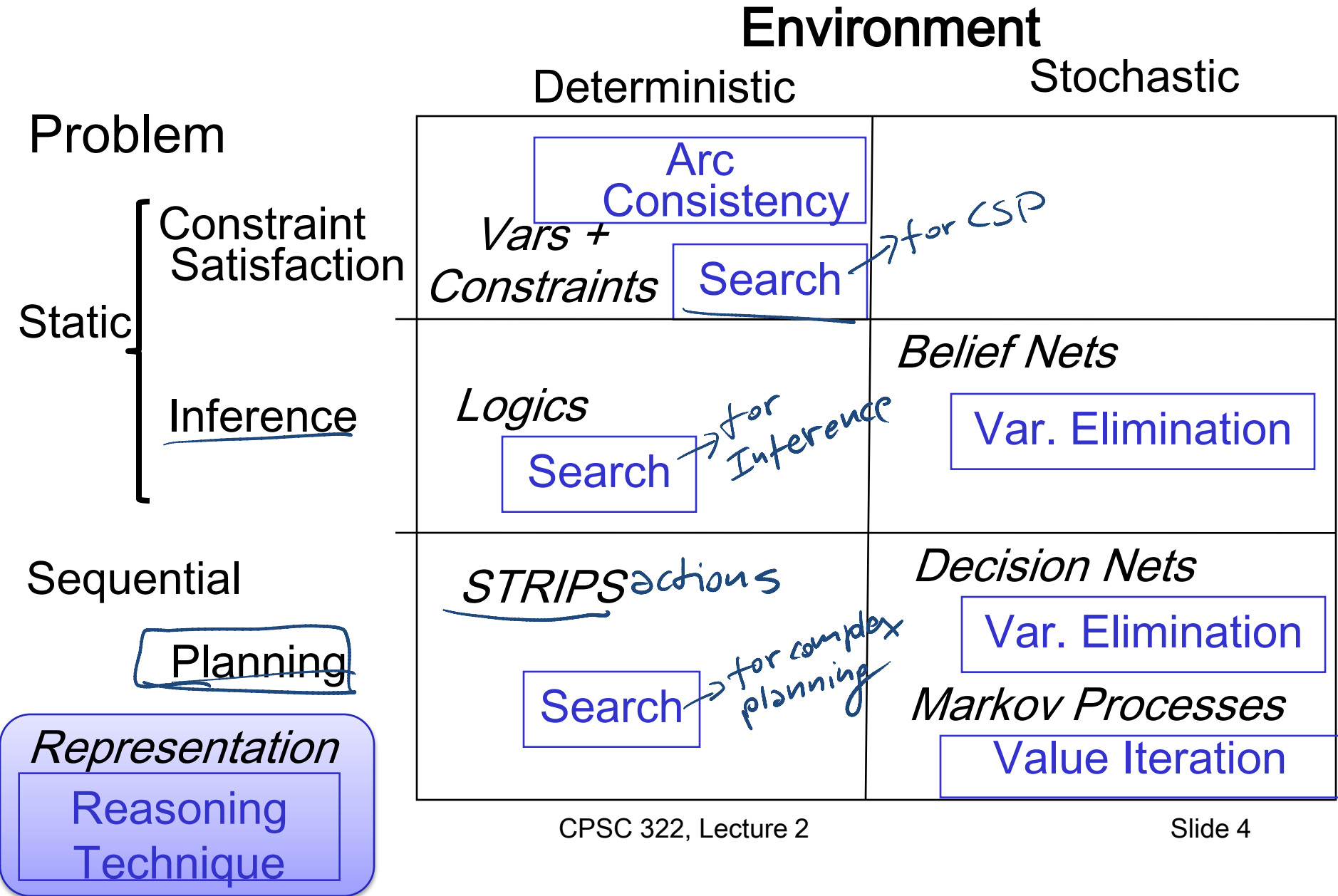
3 solutions

$A=5 \quad B=5 \quad C=1$

" "  $C=2$


" "  $C=3$

# Modules we'll cover in this course: R&Rsys



# Standard Search vs. Specific R&R systems

## Constraint Satisfaction (Problems):

- State
  - Successor function
  - Goal test
  - Solution
  - Heuristic function
- 

## Planning :

- State
- Successor function
- Goal test
- Solution
- Heuristic function

## Inference

- State
- Successor function
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# Generate-and-Test Algorithm

- **Algorithm:**

- **Generate** possible worlds one at a time
- **Test** them to see if they violate any constraints

$\text{dom } A = \{1, 2, 3, 4, 5\}$   
 $\text{dom } B = \{1, 2, 3, 4, 5\}$   
 $\text{dom } C = \{1, 2, 3\}$

For a in domA

For b in domB

For c in domC

if  $(a, b, c)$  satisfies all the constraints

return  $(a, b, c)$

return fail

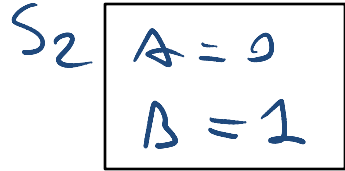
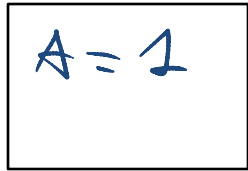
- This procedure is able to solve any CSP
- However, the running time is proportional to the number of possible worlds
  - always exponential in the number of variables
  - far too long for many CSPs ☹

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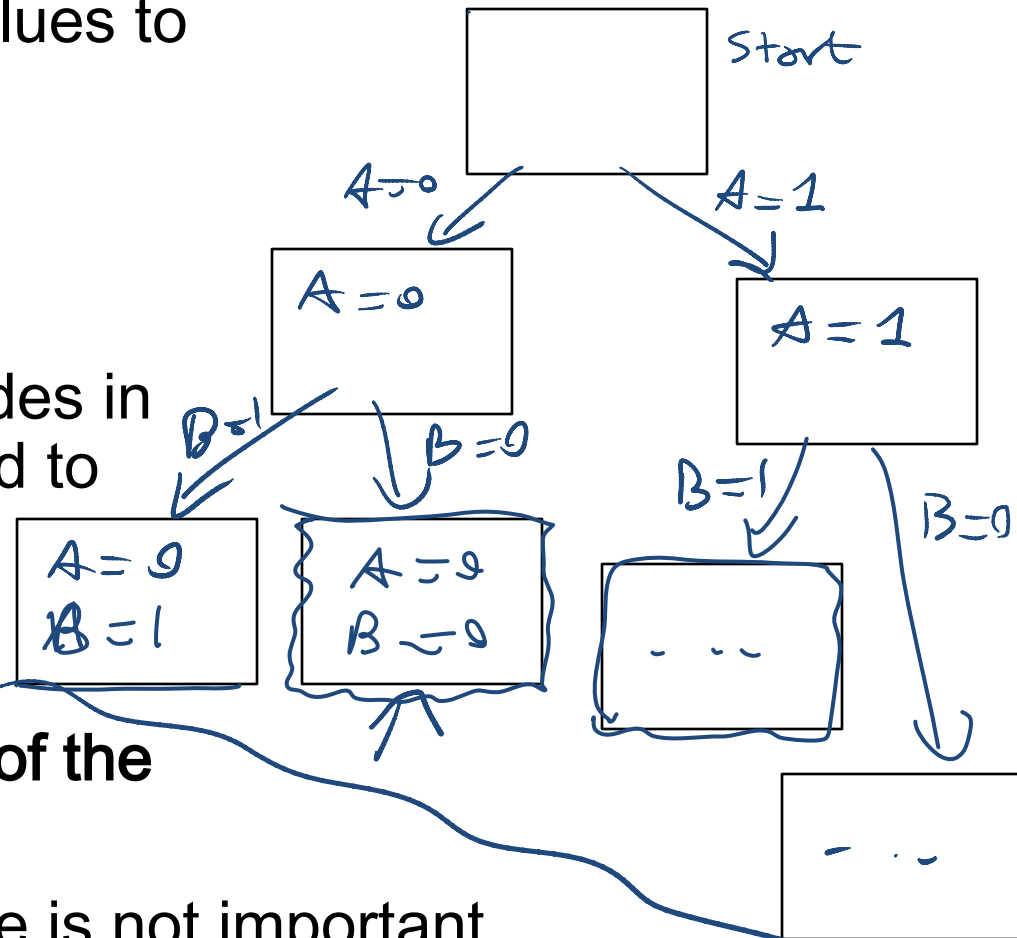
S<sub>1</sub>



# CSPs as search problems

$A, B$   $\text{dom } A = \text{dom } B = \{0, 1\}$   
 $A = B$

- **states:** assignments of values to a subset of the variables
- **start state:** the empty assignment (no variables assigned values)
- **neighbours** of a state: nodes in which values are assigned to one additional variable
- **goal state:** a state which assigns a value to each variable, and satisfies all of the constraints



Note: the path to a goal node is not important

# CSPs as Search Problems

What **search strategy** will work well for a CSP?

- there's no role for a heuristic function. If there are  $n$  variables every solution is at depth... $n$ ...

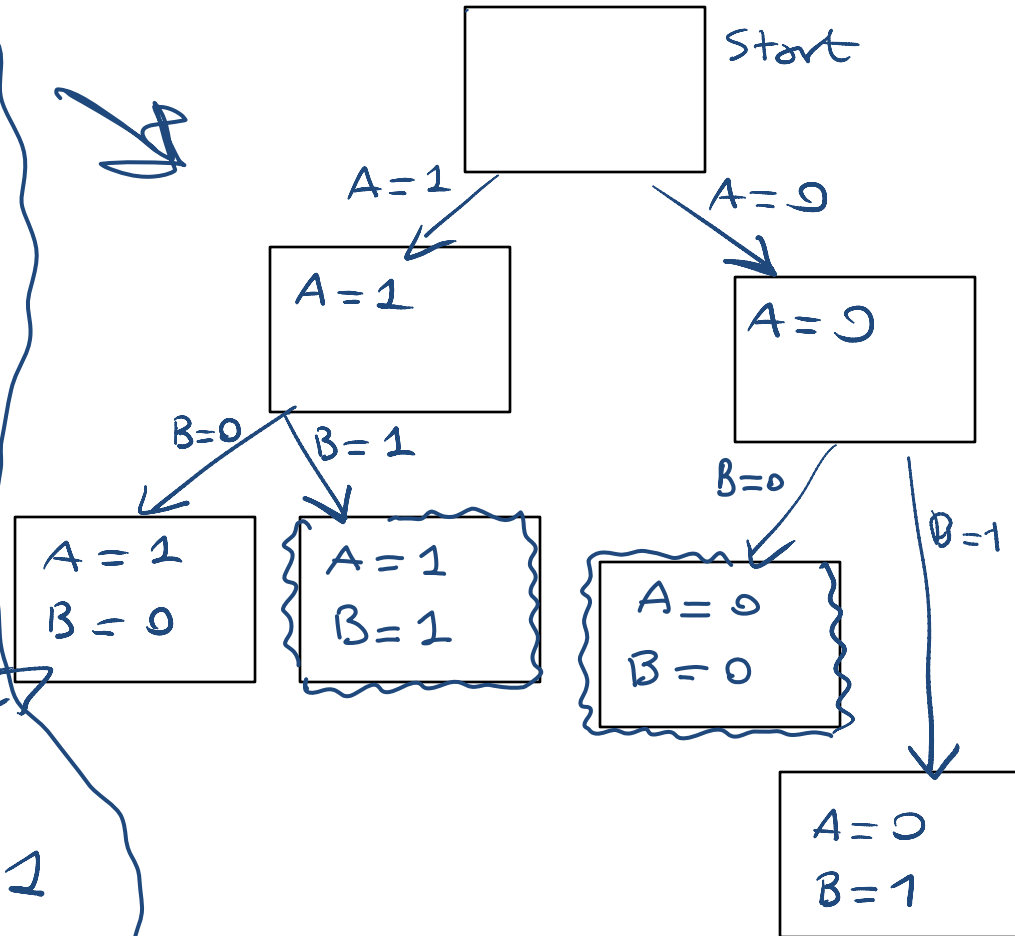
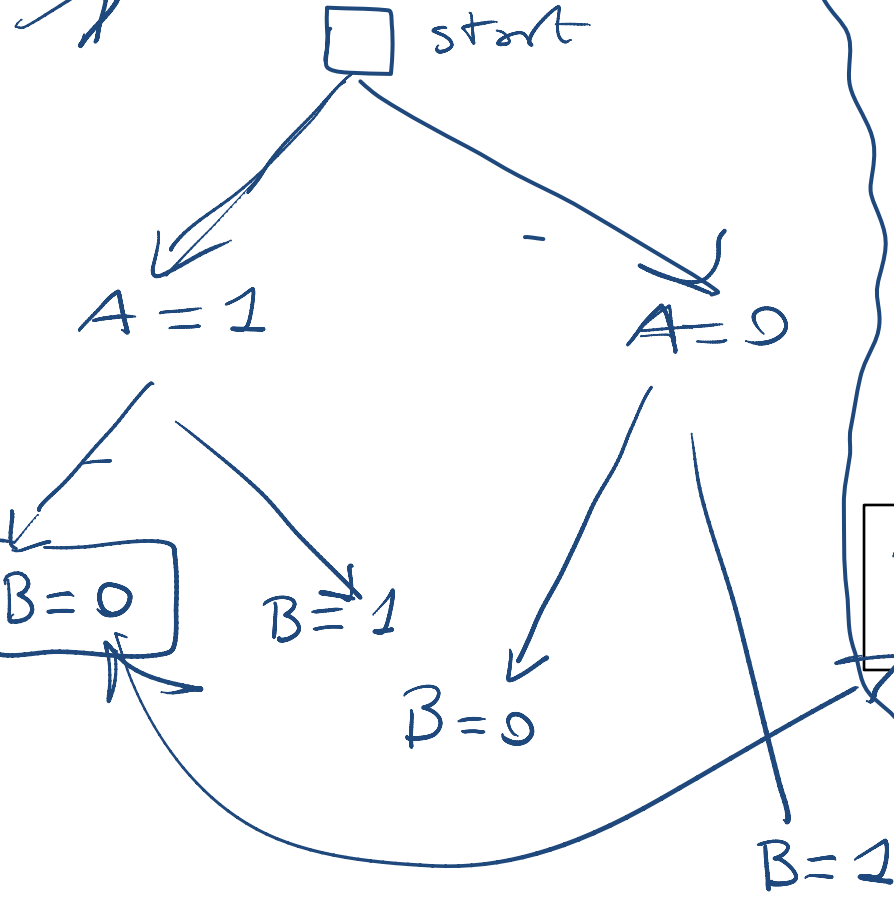
well formed

- the tree is always finite and has no cycles, so which one is better BFS or IDS or DFS?

# CSPs as search problems

$A, B \quad \text{dom } A = \text{dom } B = \{0, 1\}$   
 $\text{const } A = B$

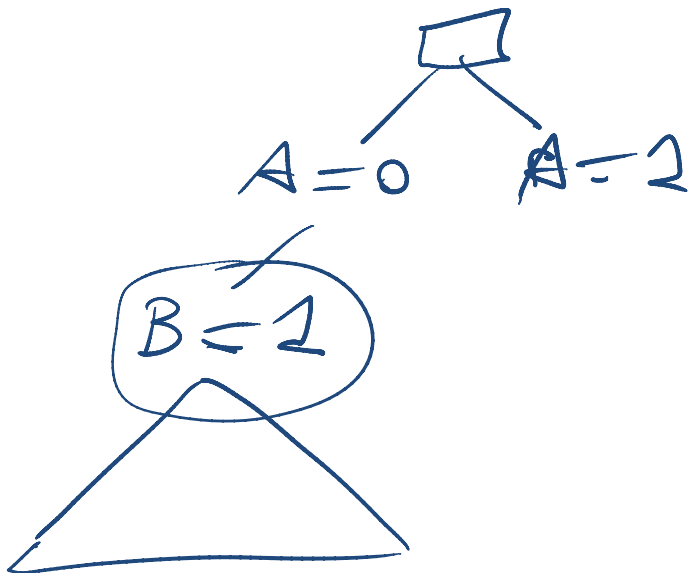
Simplified notation



# CSPs as Search Problems

How can we avoid exploring some sub-trees i.e., **prune** the DFS Search tree?

- once we consider a path whose end node violates one or more constraints, we know that a solution cannot exist below that point
- thus we should **remove that path** rather than continuing to search



down of  
all vars

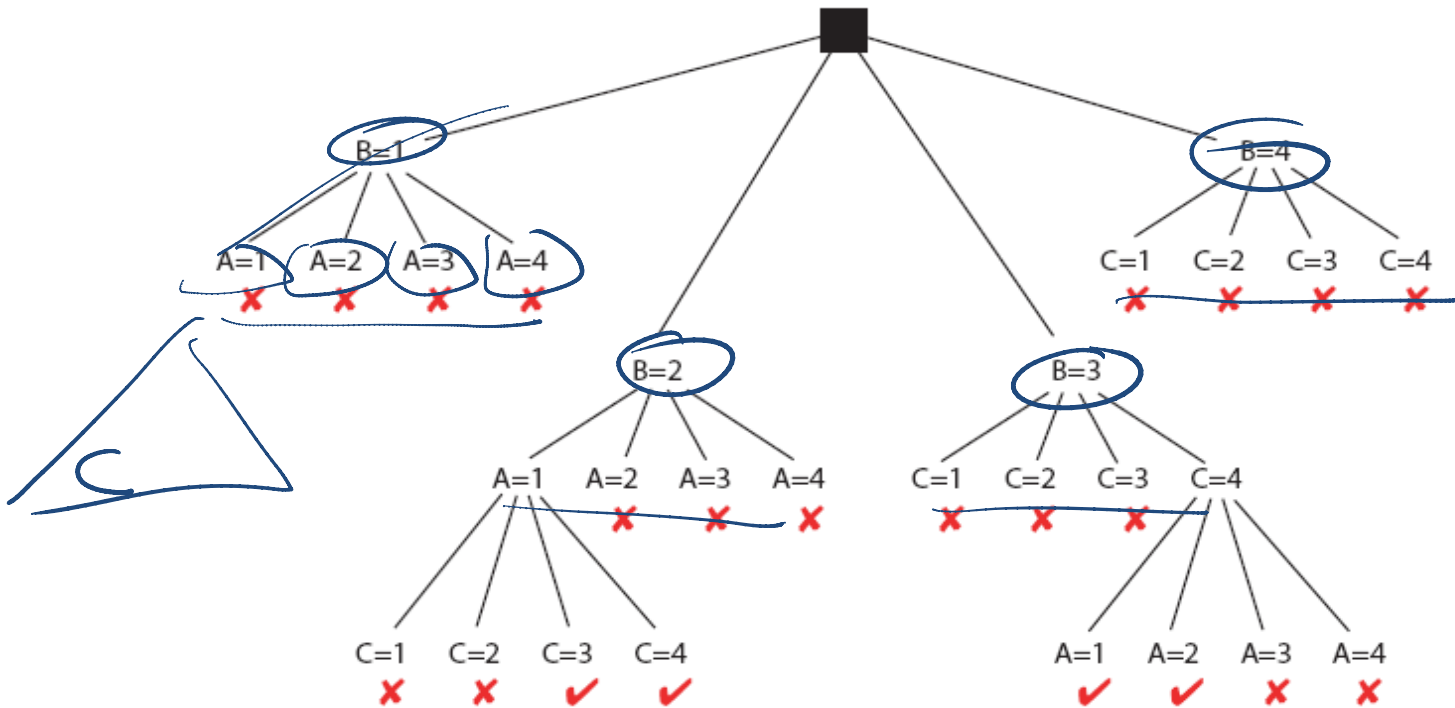
A B C D E F {1,0}

constraint  $(A = B)$

# Solving CSPs by DFS: Example

Problem:

- Variables: A,B,C
- Domains: {1, 2, 3, 4}
- Constraints: A < B, B < C



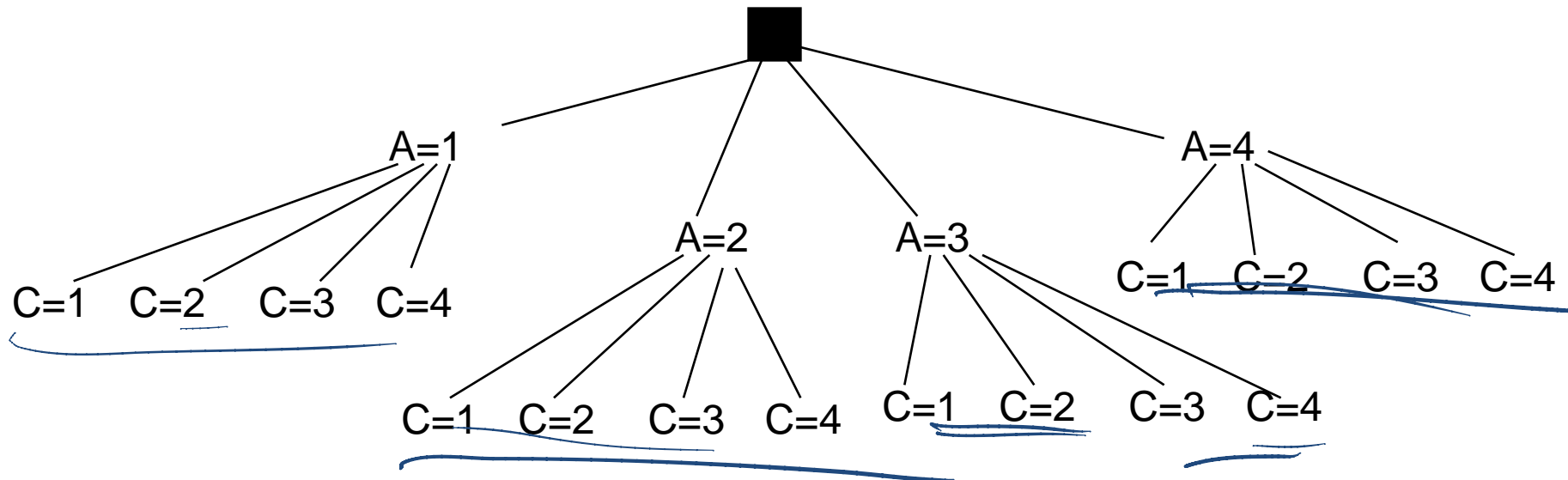
# Solving CSPs by DFS: Example Efficiency

## Problem:

- Variables: A,B,C
- Domains: {1, 2, 3, 4}
- Constraints:  $A < B$ ,  $B < C$

Note: the algorithm's efficiency depends on the order in which variables are expanded

## *Degree Heuristics*



# Standard Search vs. Specific R&R systems

## Constraint Satisfaction (Problems):

- **State:** assignments of values to a subset of the variables ↙
- **Successor function:** assign values to a “free” variable
- **Goal test:** set of constraints
- **Solution:** possible world that satisfies the constraints
- **Heuristic function:** *none (all solutions at the same distance from start)*

## Planning :

- State
- Successor function
- Goal test
- Solution
- Heuristic function

## Inference

- State
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# Can we do better than Search?

## Key ideas:

- prune the domains as much as possible before “searching” for a solution.

Simple when using constraints involving single variables  
(technically enforcing **domain consistency**)

**Definition:** A variable is **domain consistent** if no value of its domain is ruled impossible by any unary constraints.

- Example:  $D_B = \{1, 2, \cancel{3}, 4\}$  *is not* domain consistent if we have the constraint  $B \neq 3$ .

# How do we deal with constraints involving multiple variables?

Definition (constraint network)

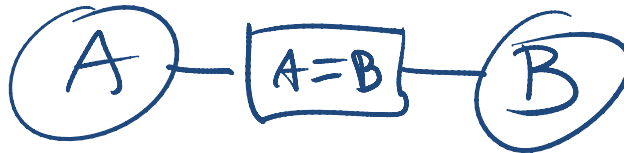
A **constraint network** is defined by a graph, with

- one node for every variable
- one node for every constraint

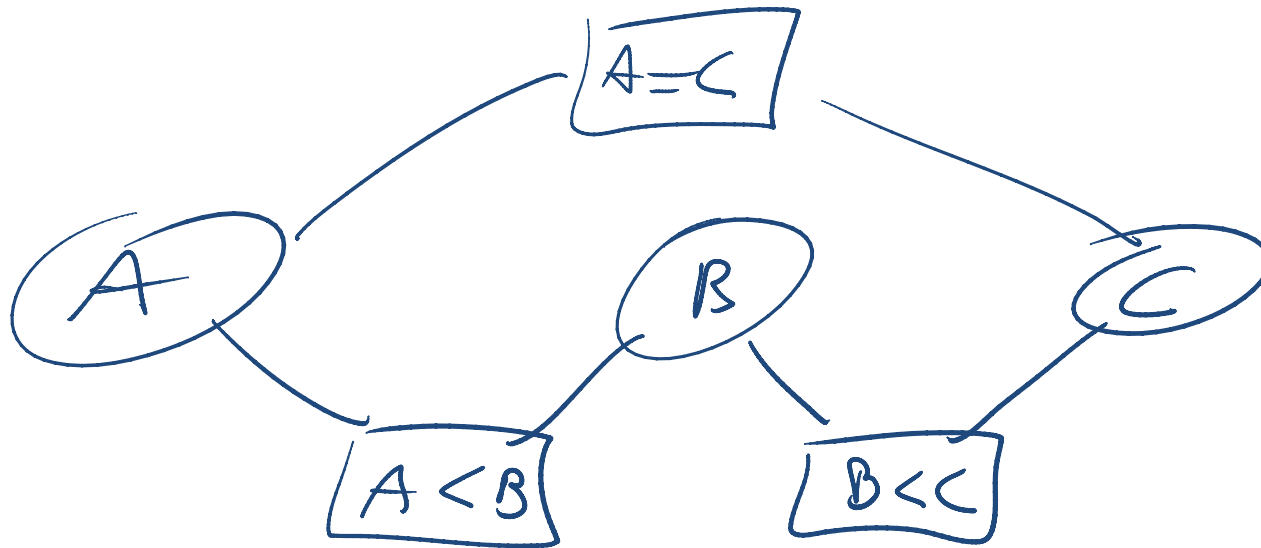
and undirected edges running between variable nodes and constraint nodes whenever a given variable is involved in a given constraint.

$A \quad B \quad \{1, 0\}$

$A = B$



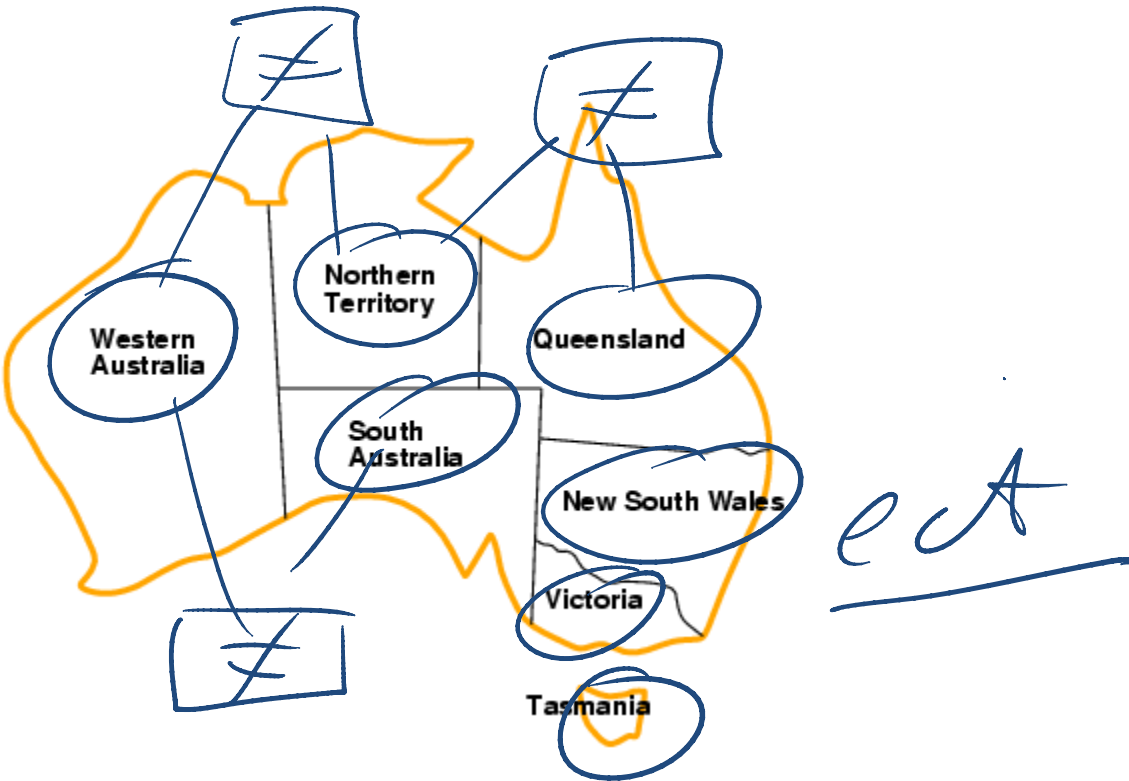
# Example Constraint Network



Recall:

- Variables: A,B,C
- Domains: {1, 2, 3, 4}
- Constraints:  $A < B$ ,  $B < C$ ,  $C = A$

# Example: Constraint Network for Map-Coloring



**Variables**  $WA, NT, Q, NSW, V, SA, T$

**Domains**  $D_i = \{\text{red, green, blue}\}$

**Constraints:** adjacent regions must have different colors

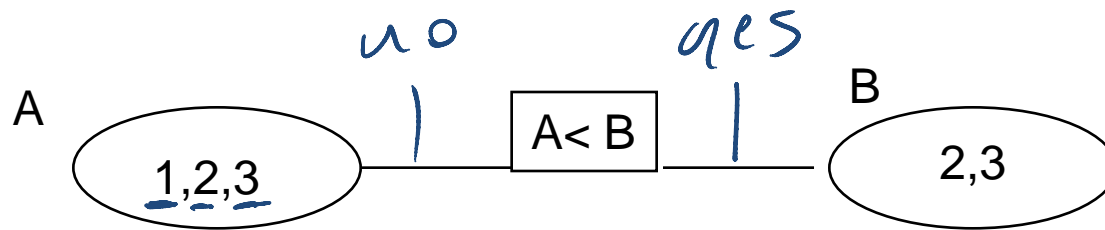
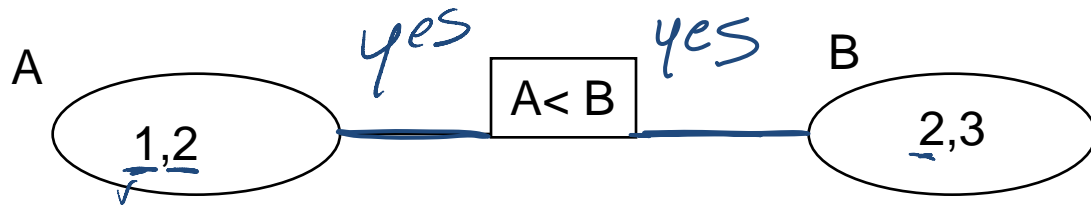
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# Arc Consistency

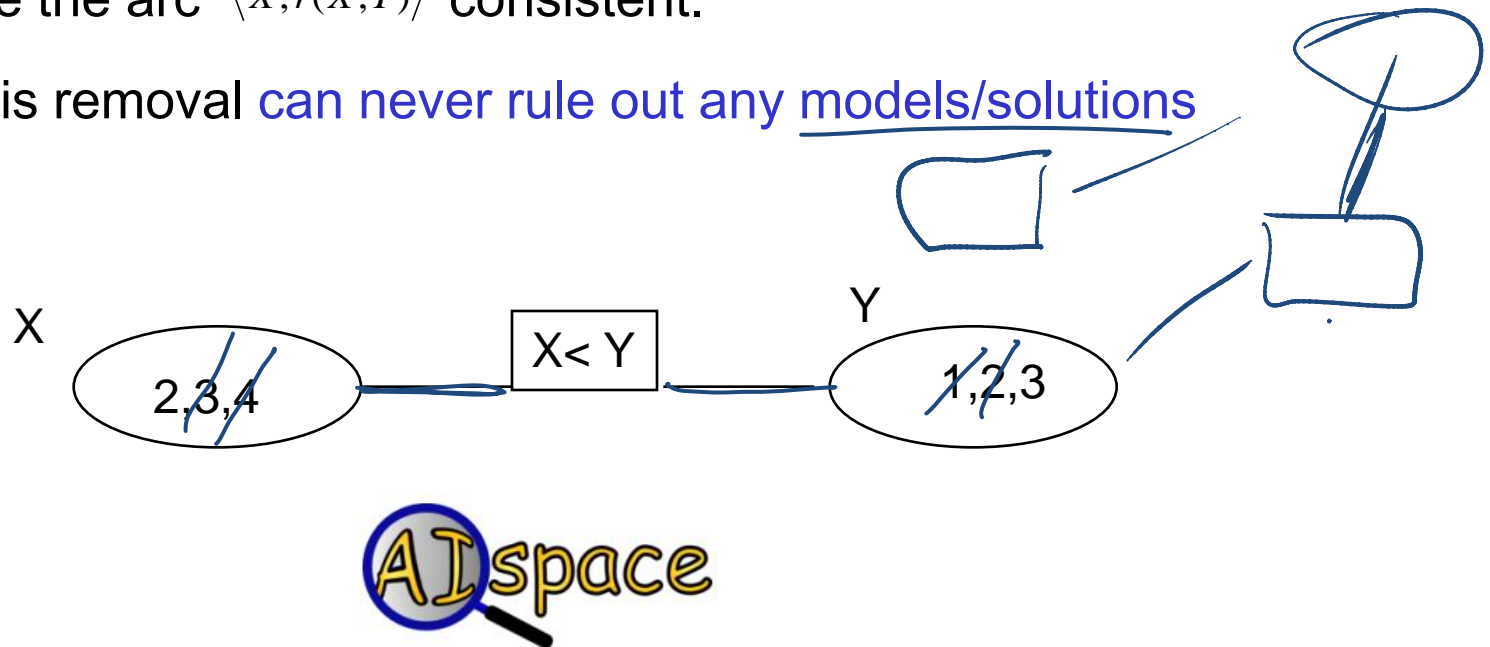
## Definition (arc consistency)

An arc  $\langle X, r(X, Y) \rangle$  is **arc consistent** if for each value  $x$  in  $\text{dom}(X)$  there is some value  $y$  in  $\text{dom}(Y)$  such that  $r(x, y)$  is satisfied.



# How can we enforce Arc Consistency?





- If an arc  $\langle X, r(X,Y) \rangle$  is not arc consistent, all values  $x$  in  $dom(X)$  for which there is no corresponding value in  $dom(Y)$  may be deleted from  $dom(X)$  to make the arc  $\langle X, r(X,Y) \rangle$  consistent.
- This removal can never rule out any models/solutions



- A network is arc consistent if all its arcs are arc consistent. 

# Learning Goals for today's class

## You can:

- Implement the **Generate-and-Test** Algorithm. Explain its disadvantages. 
- Solve a **CSP by search** (specify neighbors, states, start state, goal state). Compare strategies for CSP search. Implement pruning for **DFS** search in a CSP. 
- Build a **constraint network** for a set of constraints.
- Verify whether a network is arc consistent.   
*make an arc*  *arc-consistent*  
*only an arc today. The whole network next lecture*



## Next class

How to make a constraint network arc consistent?  
**Arc Consistency Algorithm**



**CSP Practice Exercise posted:  
check it out!**