

CSPs: Search and Arc Consistency

Computer Science cpsc322, Lecture 12

(Textbook Chpt 4.3-4.5)

Oct, 2, 2013



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

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

- a domain for each variable

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

- a set of constraints

$\begin{matrix} \hookrightarrow & B=5 & C > B & A=B & \text{no sol} \\ \hookrightarrow & \text{"} & C < B & \text{"} & \end{matrix}$

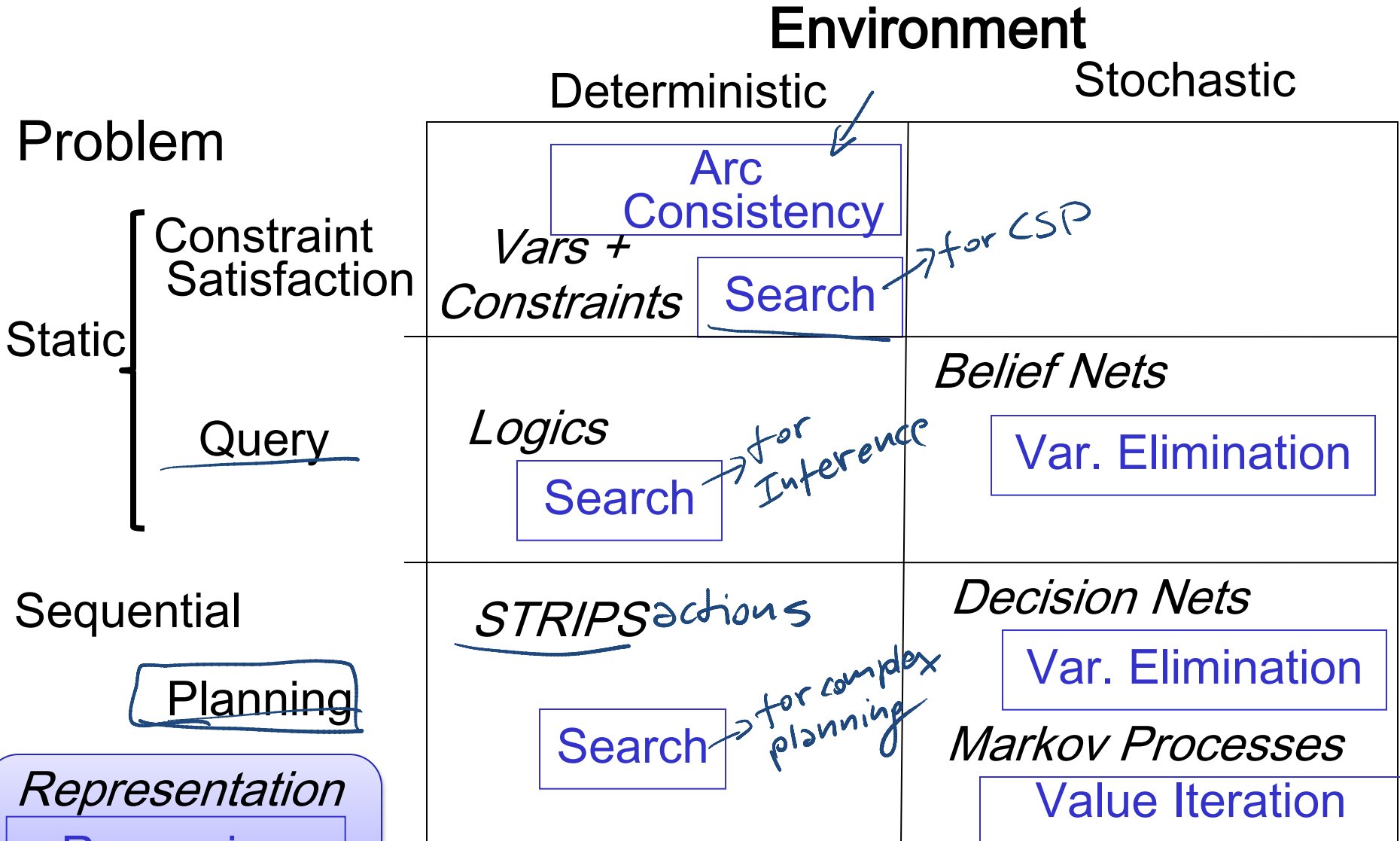
possible worlds
3 * 5 * 5

Definition (model / solution)

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

$\begin{matrix} \rightarrow & B=5 & A=5 & C=1 \\ \rightarrow & 1 & \text{"} & C=2 \\ \rightarrow & 2 & \text{"} & C=3 \end{matrix}$

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



Representation
Reasoning
Technique

Standard Search vs. Specific R&R systems

Constraint Satisfaction (Problems):

- State *& start state*
- Successor function
- Goal test
- Solution
- Heuristic function

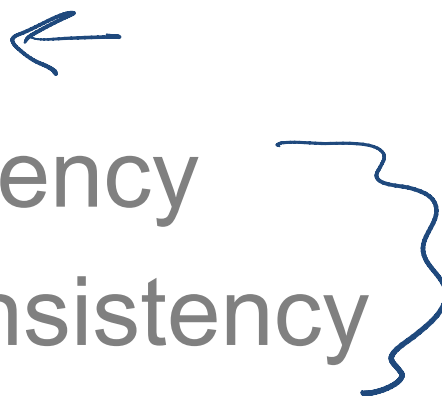
Planning :

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

Answering Queries

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

Lecture Overview

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

Generate-and-Test Algorithm

- **Algorithm:**

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

→ dom A = {1, 2, 3, 4, 5}
→ dom B = {1, 2, 3, 4, 5}
→ dom C = {1, 2, 3}

```
For a in domA
```

```
  For b in domB
```

```
    For c in domC
```

```
      if (abc) satisfies all constraints
```

```
        return (abc)
```

```
return fail
```

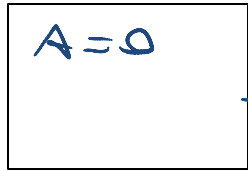
- This procedure is able to solve any CSP
- However, the running time is proportional to the number of possible worlds
 - always exponential in the number of variables
 - far too long for many CSPs ☹

Lecture Overview

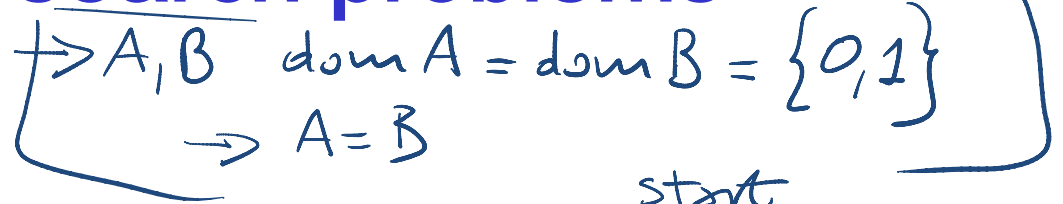
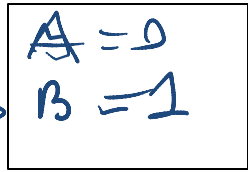
- Recap
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- **Search**
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CSPs as search problems

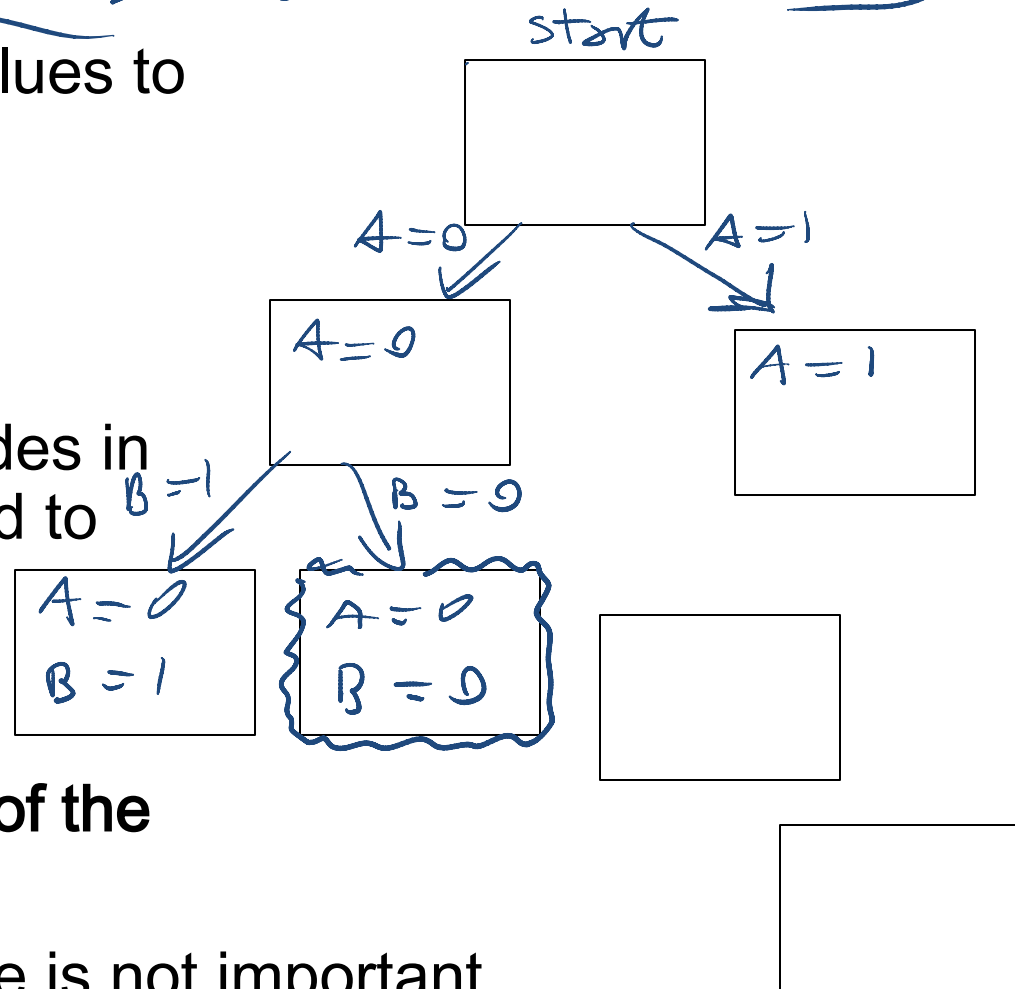
S₁



S₂



- **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?

- If there are n variables every solution is at depth..... ^{n}

Is there a role for a heuristic function?

A. Yes

B. No

C. It depends



The search space is always?

A. Finite with cycles

B. Infinite without cycles

C. Finite without cycles

D. Infinite with cycles



So which search strategy is better?

A. BFS

B. IDS

C. A*

D. DFS

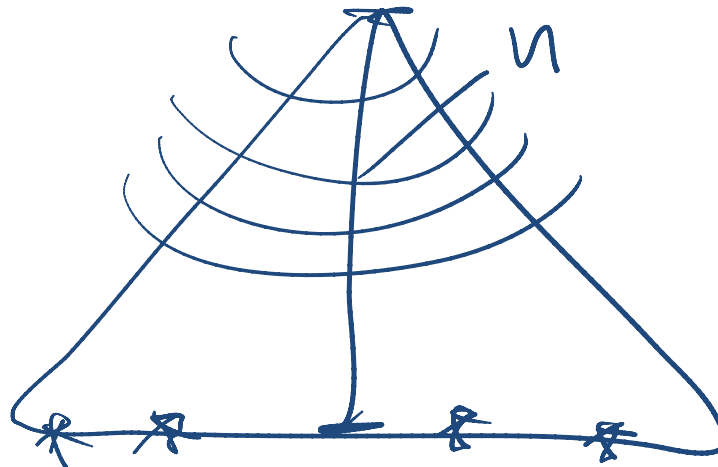


CSPs as Search Problems

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

- If there are n variables every solution is at depth n ...
- Is there a role for a heuristic function?

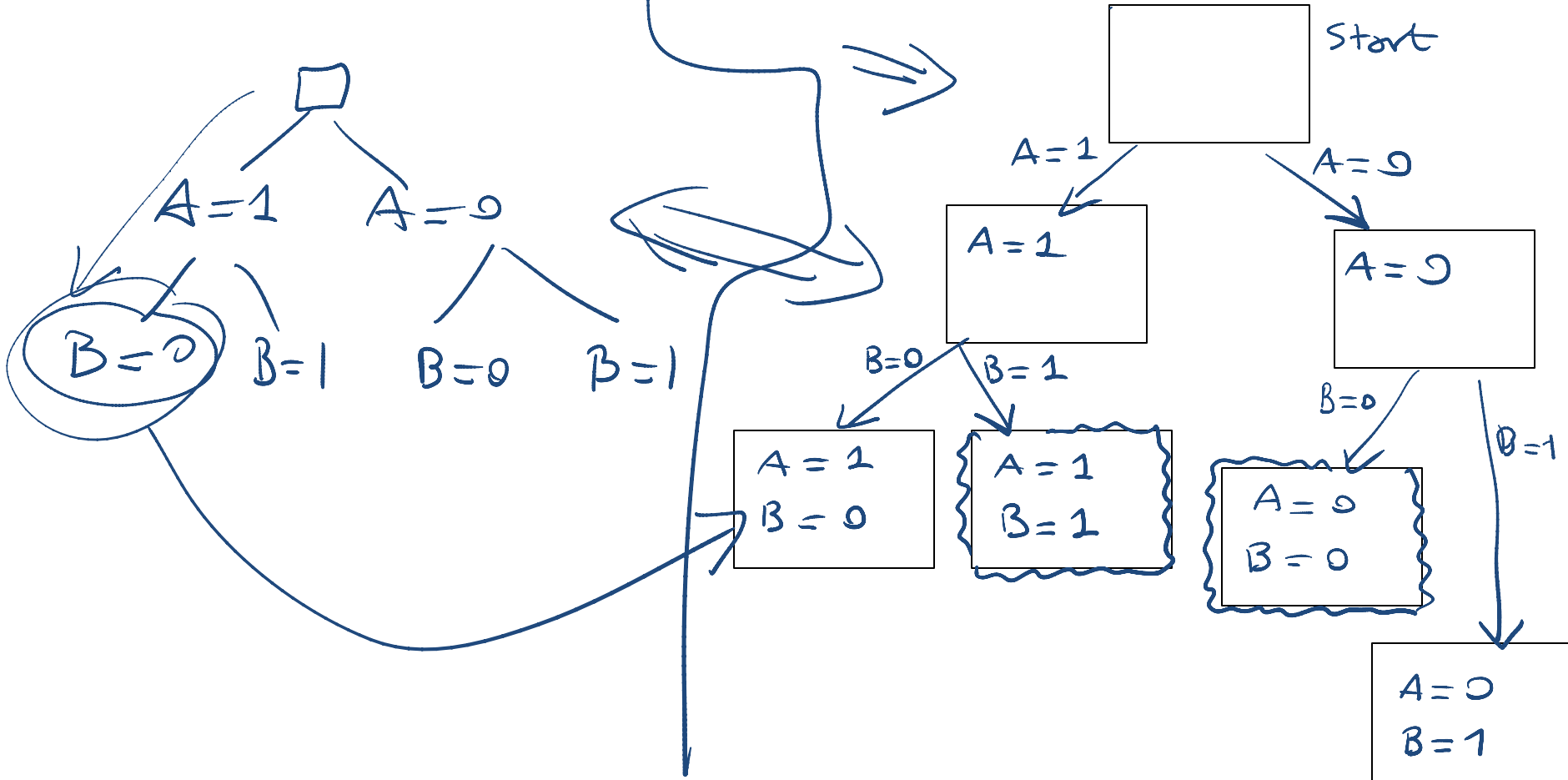
- the tree is always finite... and has no cycles..., so
which one is better ~~BFS~~ or ~~IDS~~ or DFS?



CSPs as search problems

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

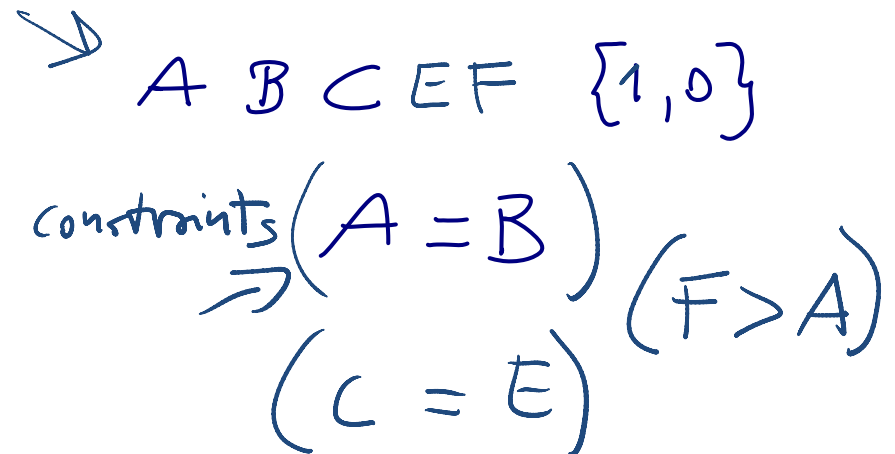
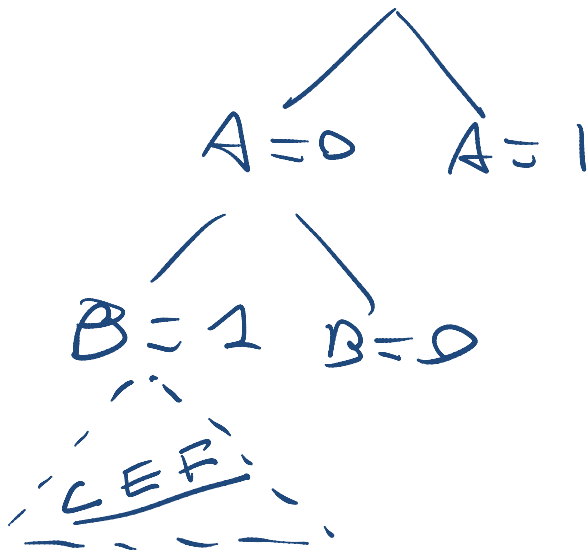
Simplified notation



CSPs as Search Problems

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

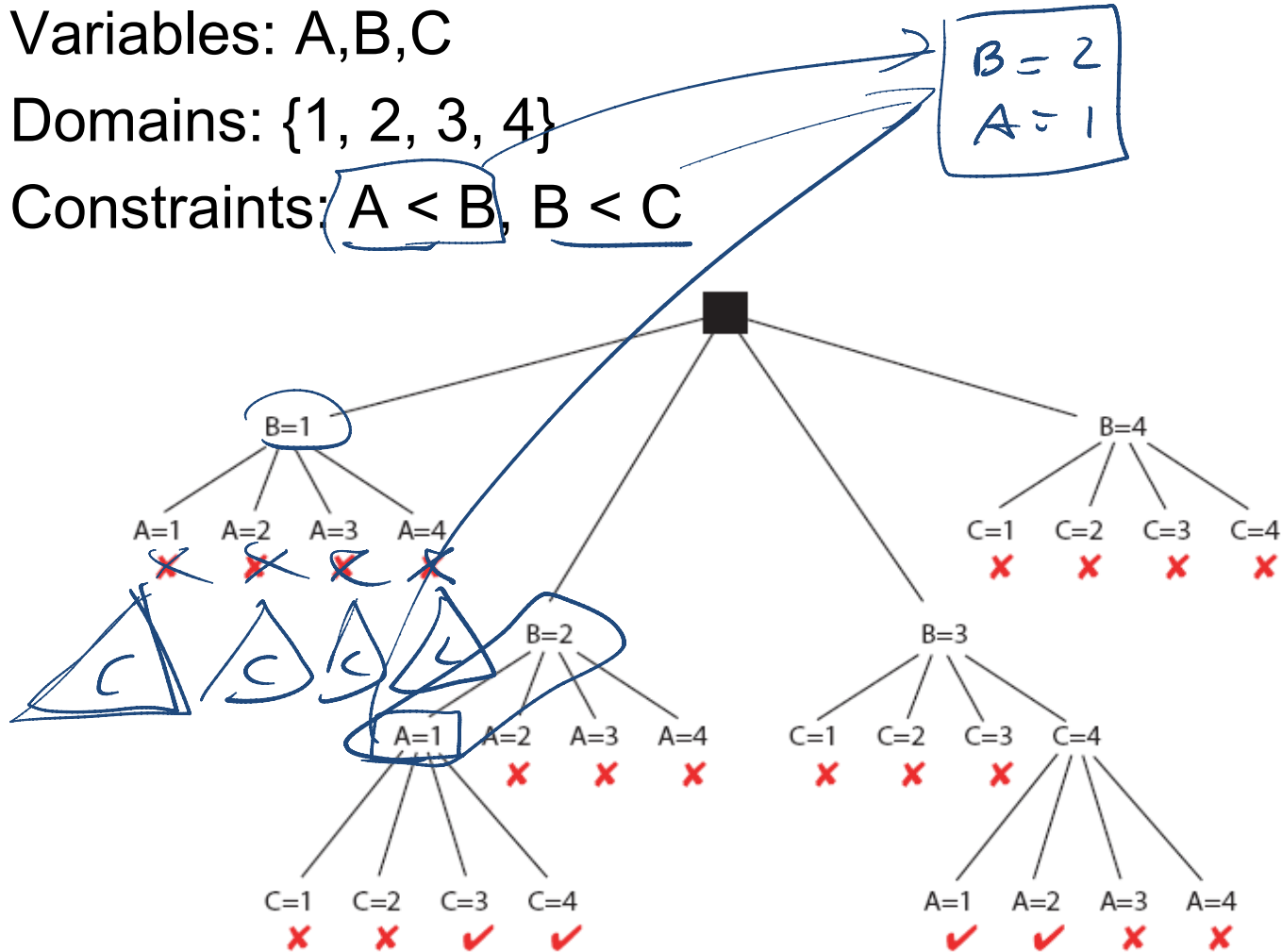
- once we consider a path whose end node violates one or more constraints, we know that a solution cannot exist below that point
- thus we should **remove that path** rather than continuing to search



Solving CSPs by DFS: Example

Problem:

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



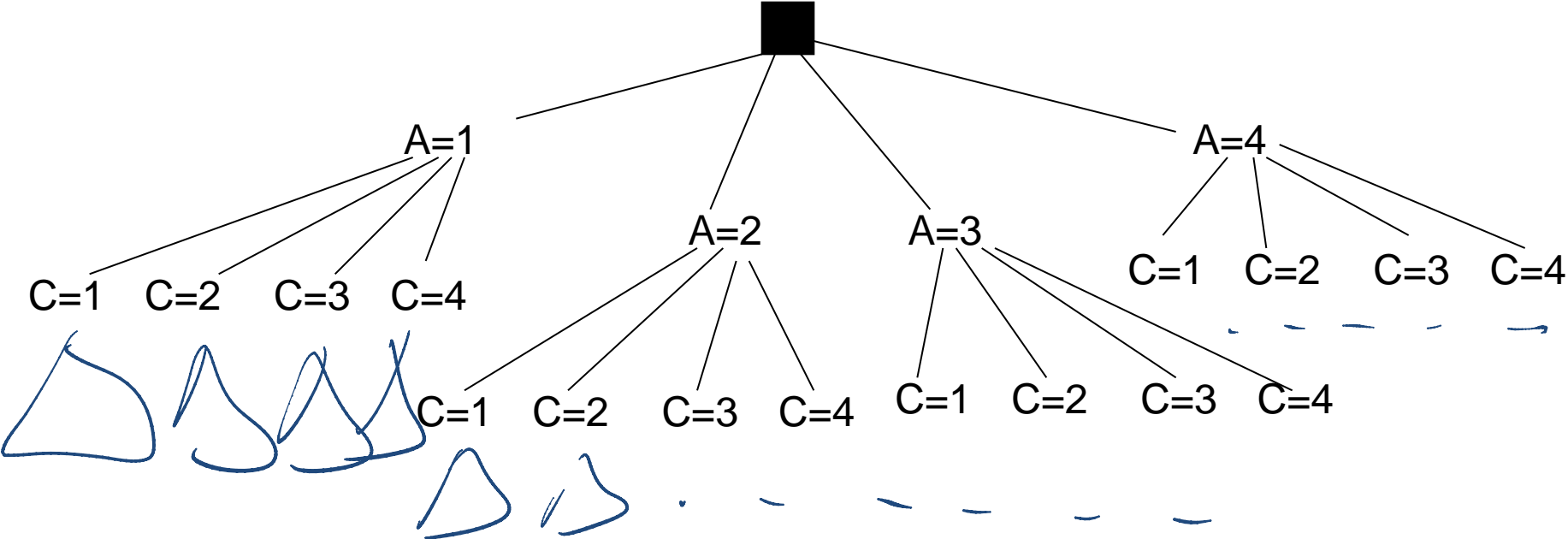
Solving CSPs by DFS: Example Efficiency

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
- Successor function
- Goal test
- Solution
- Heuristic function

Lecture Overview

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

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

to make it consistent

How do we deal with constraints involving multiple variables?

Definition (constraint network)

A constraint network is defined by a graph, with

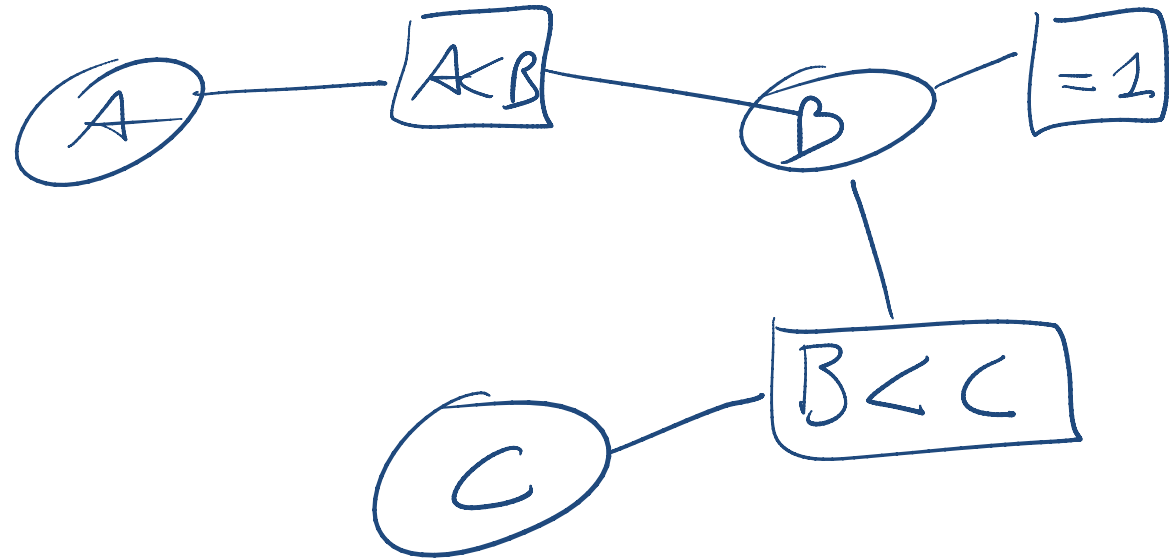
- one node for every variable
- one node for every constraint

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

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



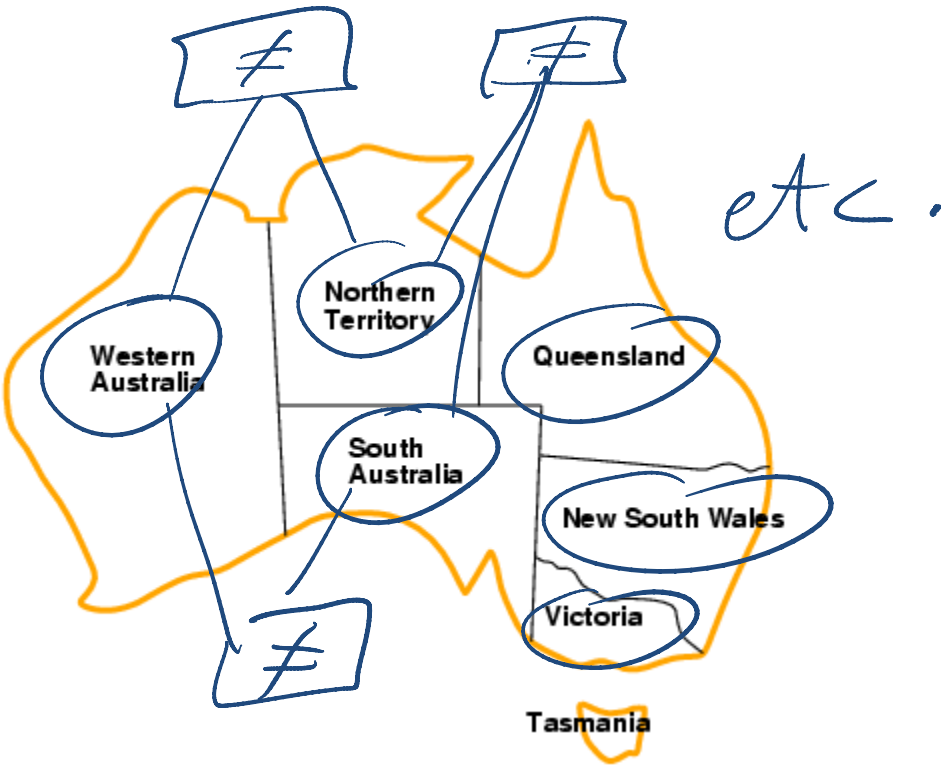
Example Constraint Network



Recall Example:

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

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

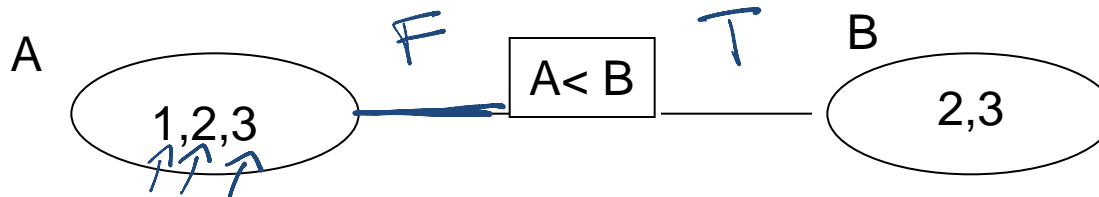
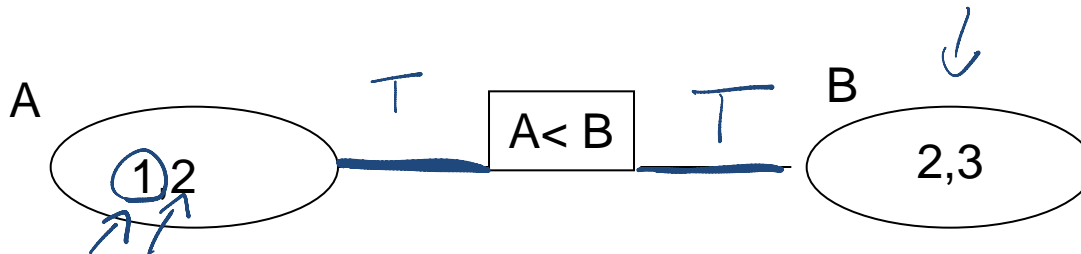
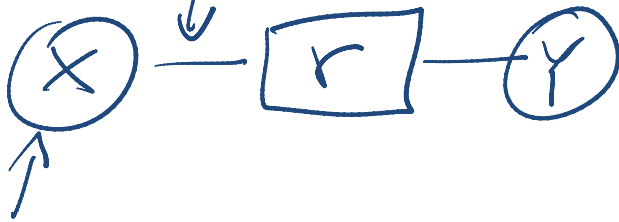
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- **Arc Consistency**

Arc Consistency

Definition (arc consistency)

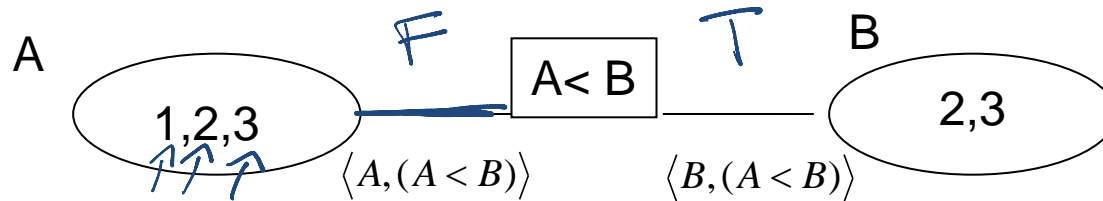
An arc $\langle X, r(X, Y) \rangle$ is **arc consistent** if for each value x in $\text{dom}(X)$ there is some value y in $\text{dom}(Y)$ such that $r(x, y)$ is satisfied.



Arc Consistency

Definition (arc consistency)

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

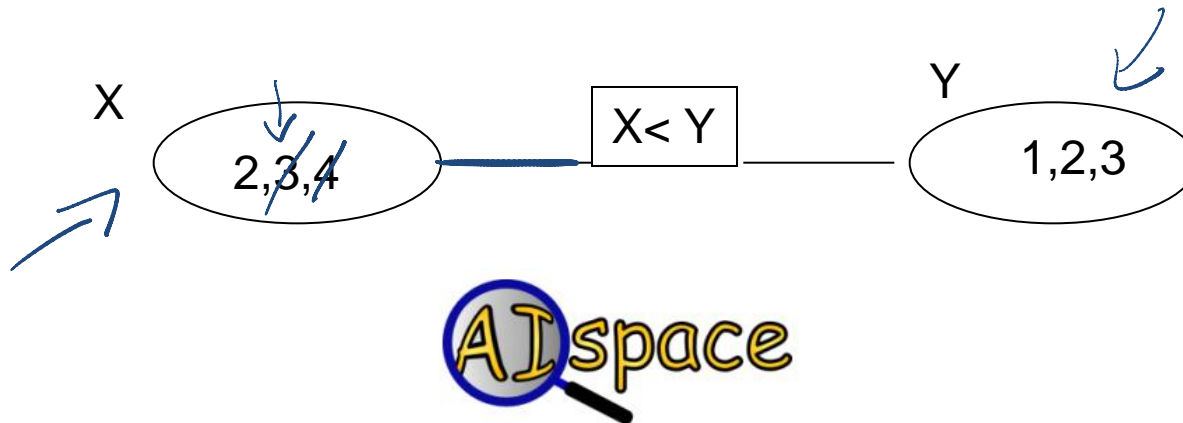


- A. Both arcs are consistent
- B. Left consistent, right inconsistent
- C. Right ~~in~~ inconsistent, left consistent
- D. Both arcs are inconsistent



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. ↵

Next class

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

There are Practice Exercises for CSP