

CSPs: Search and Arc Consistency

Computer Science cpsc322, Lecture 12

(Textbook Chpt 4.3-4.5)

January, 29, 2010

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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 ^{# possible worlds}
 $3 \times 5 \times 5$

- a set of variables

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

- a domain for each variable

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

- a set of constraints

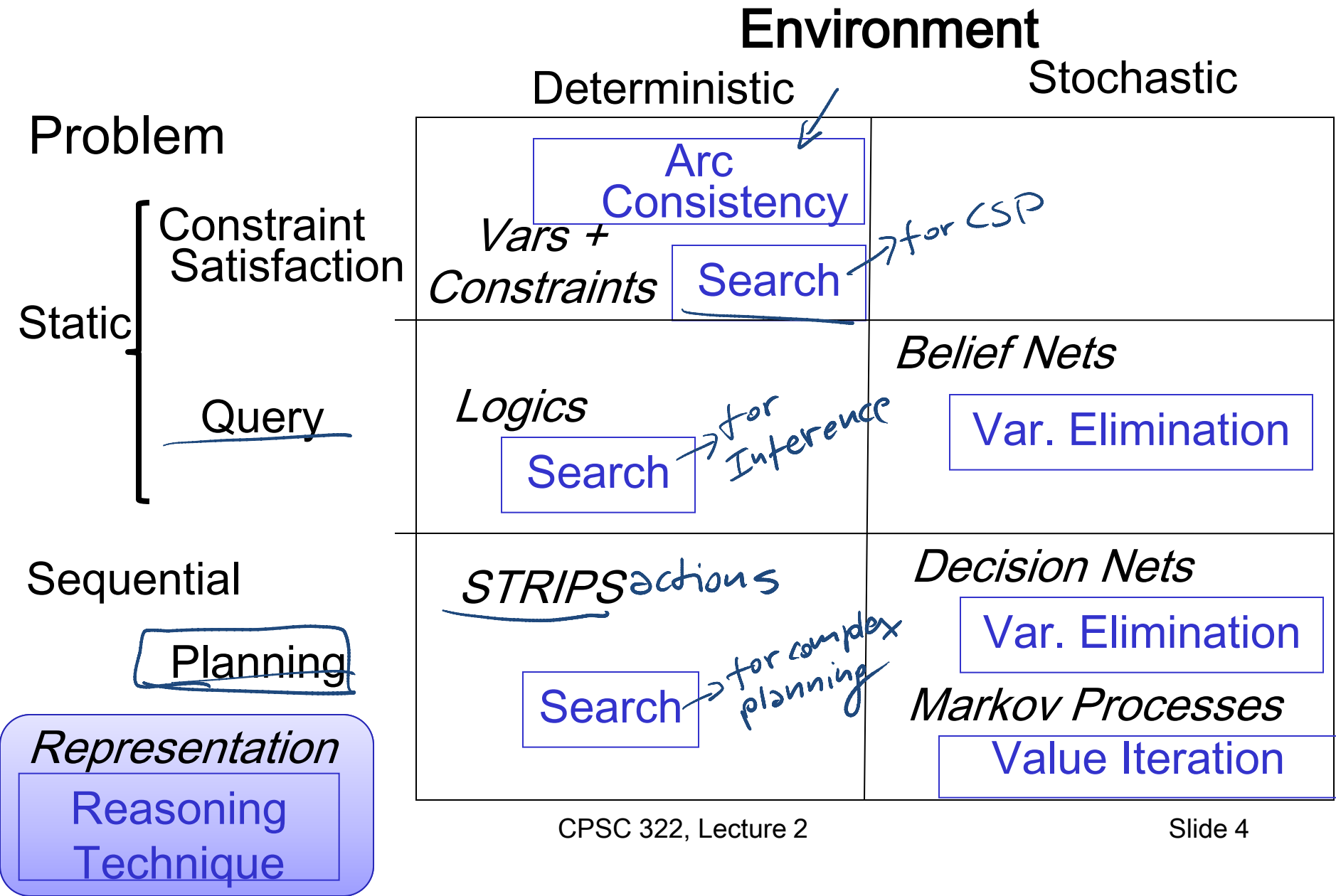
$$\begin{array}{l} \rightarrow B=5 \quad C > B \quad A=B \quad \text{no sol} \\ \rightarrow \quad \quad C < B \quad \quad \quad \checkmark \end{array}$$

Definition (model / solution)

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

$$\begin{array}{l} \rightarrow B=5 \quad A=5 \quad C=1 \\ \rightarrow \quad \quad \quad \quad \quad \quad \quad \quad \quad C=2 \\ \rightarrow \quad \quad \quad \quad \quad \quad \quad \quad \quad C=3 \end{array}$$

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



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

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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  $(abc)$  satisfies all constraints
```

```
        return  $(abc)$ 
```

```
  return fail
```

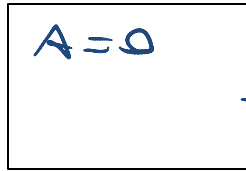
- This procedure is able to solve any CSP
- However, the running time is proportional to the number of possible worlds
 - always exponential in the number of variables
 - far too long for many CSPs ☹

Lecture Overview

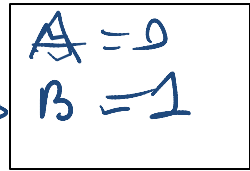
- Recap
- Generate-and-Test
- **Search**
- Consistency
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CSPs as search problems

S₁

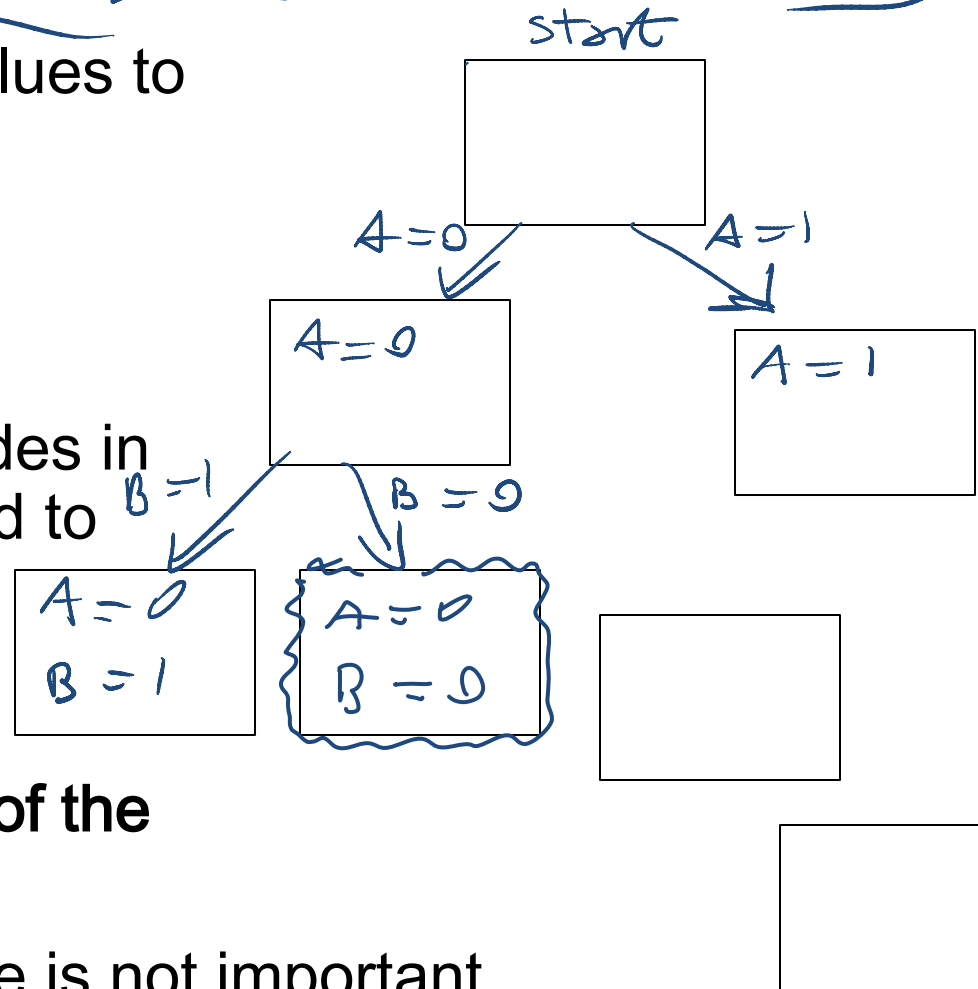


S₂



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

- **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



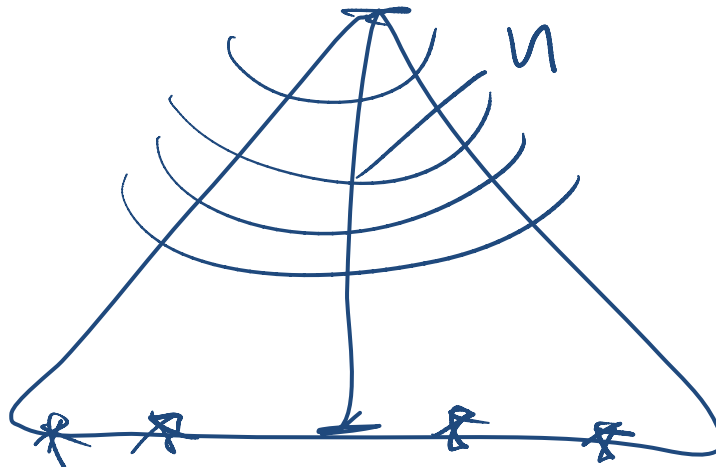
Note: the **path** to a goal node is not important

CSPs as Search Problems

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

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

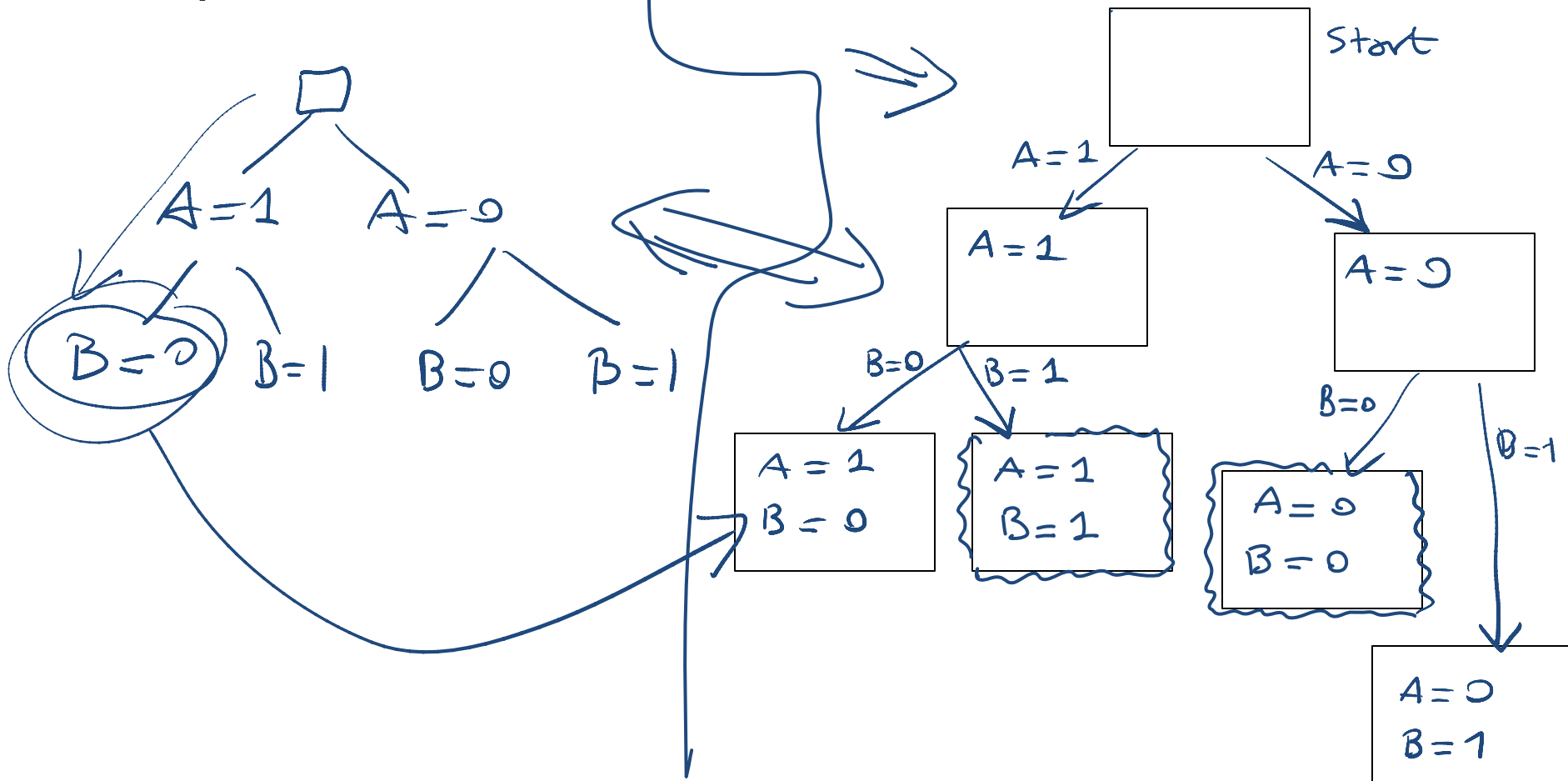
- the tree is always finite and has no cycles, so which one is better ~~BFS~~ or ~~IDS~~ or DFS?



CSPs as search problems

Simplified notation

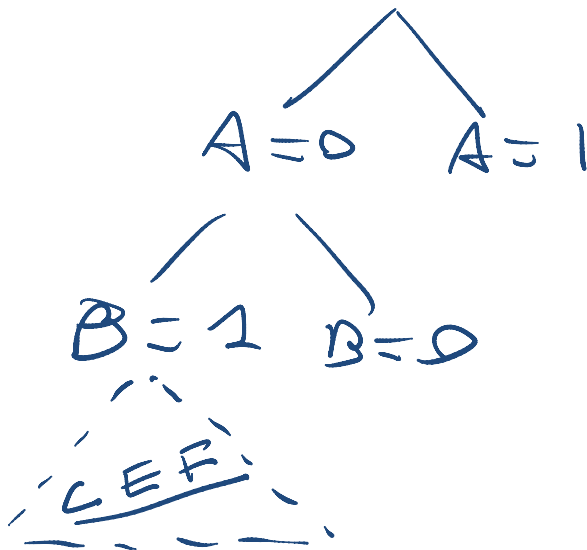
$A, B \quad \text{dom } A = \text{dom } B = \{0, 1\}$
 $\text{const } A = B$



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



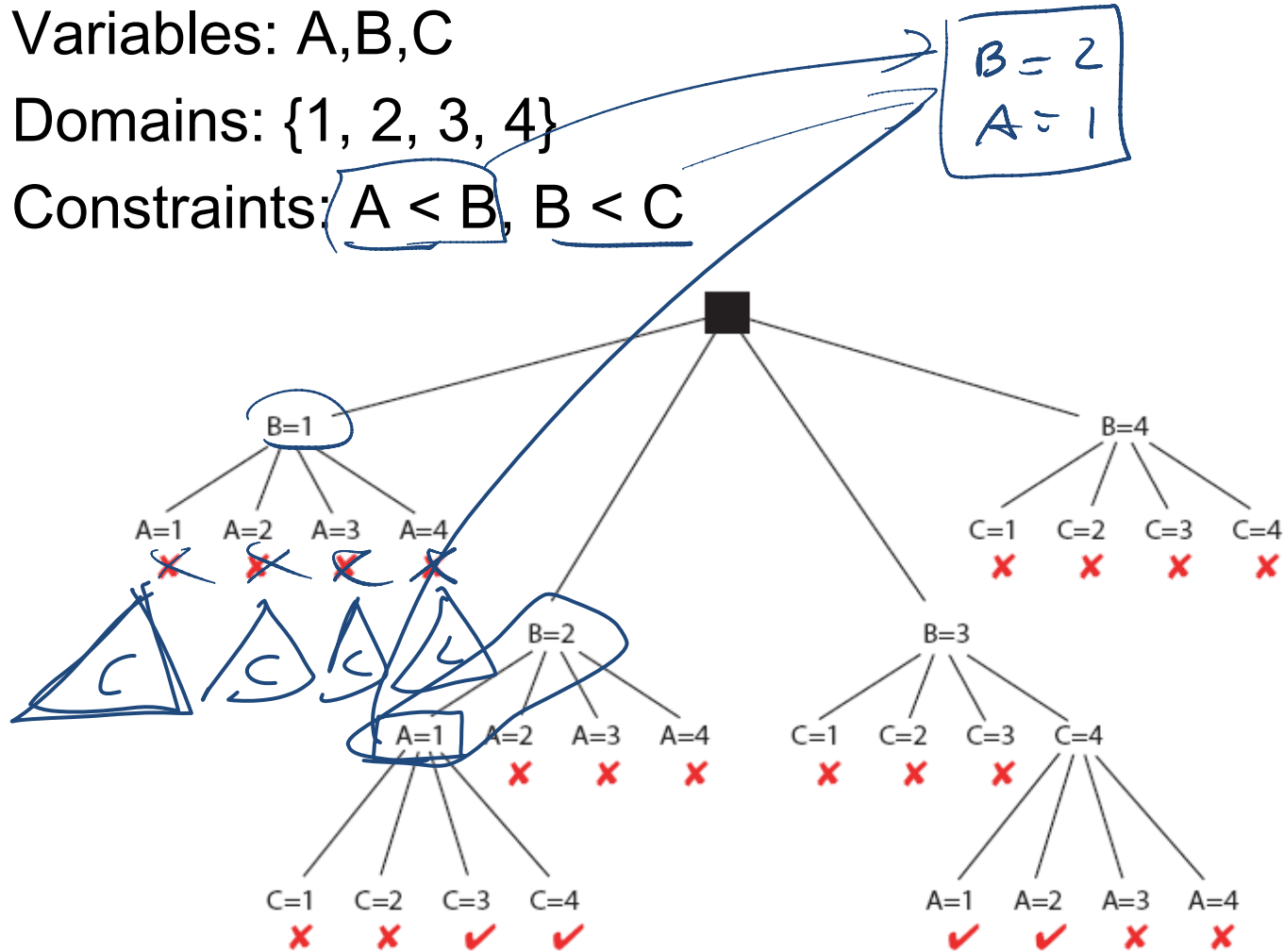
$\rightarrow A \ B \ C \ E \ F \ \{1,0\}$

$\text{constraints} \left(\begin{matrix} A = B \\ C = E \end{matrix} \right) (F > A)$

Solving CSPs by DFS: Example

Problem:

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



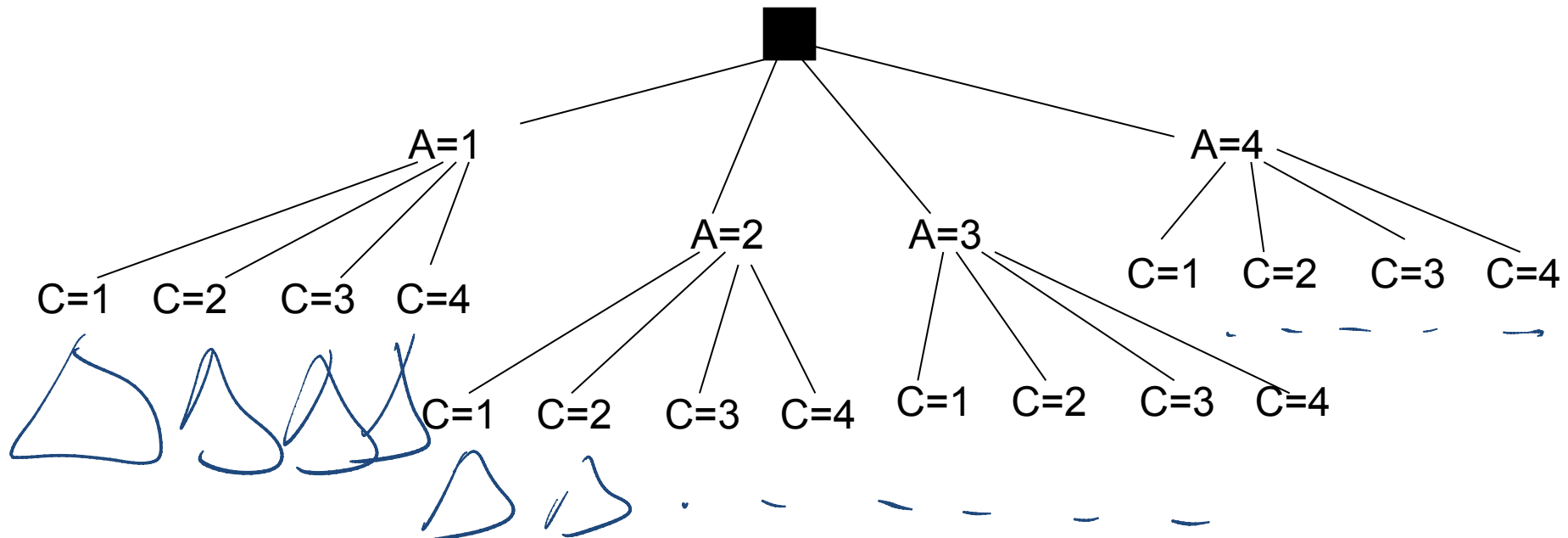
Solving CSPs by DFS: Example Efficiency

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: $D_B = \{1, 2, \cancel{3}, 4\}$ *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

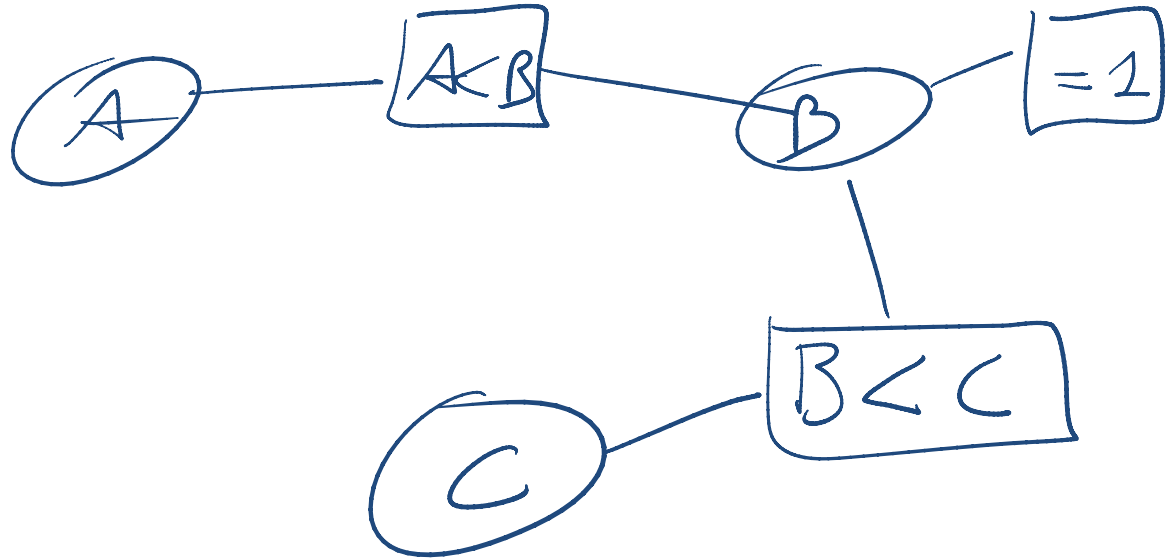
- one node for every variable
- one node for every constraint

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

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



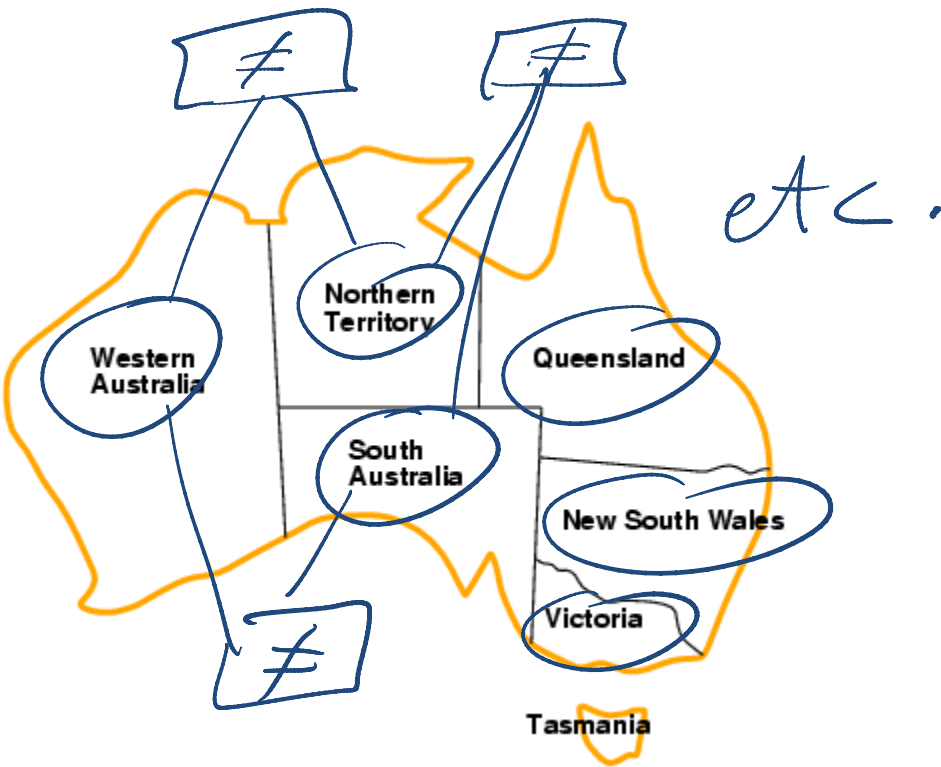
Example Constraint Network



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

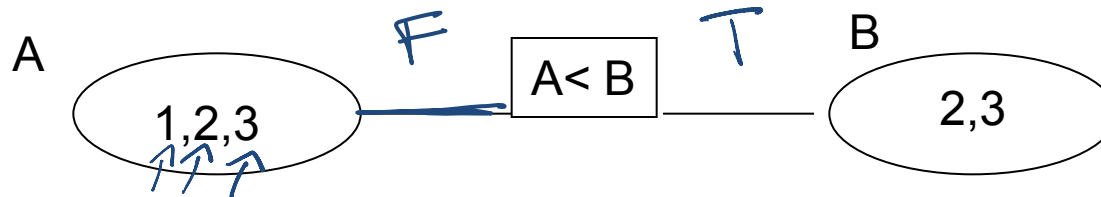
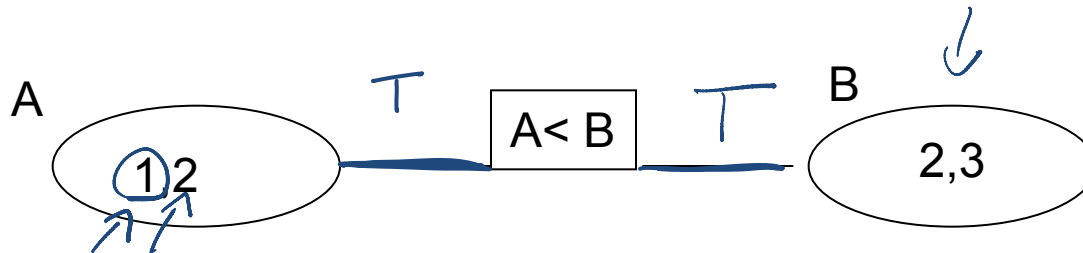
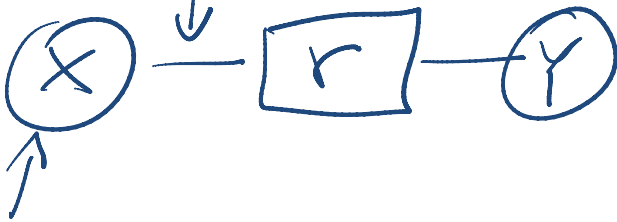
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- **Arc Consistency**

Arc Consistency

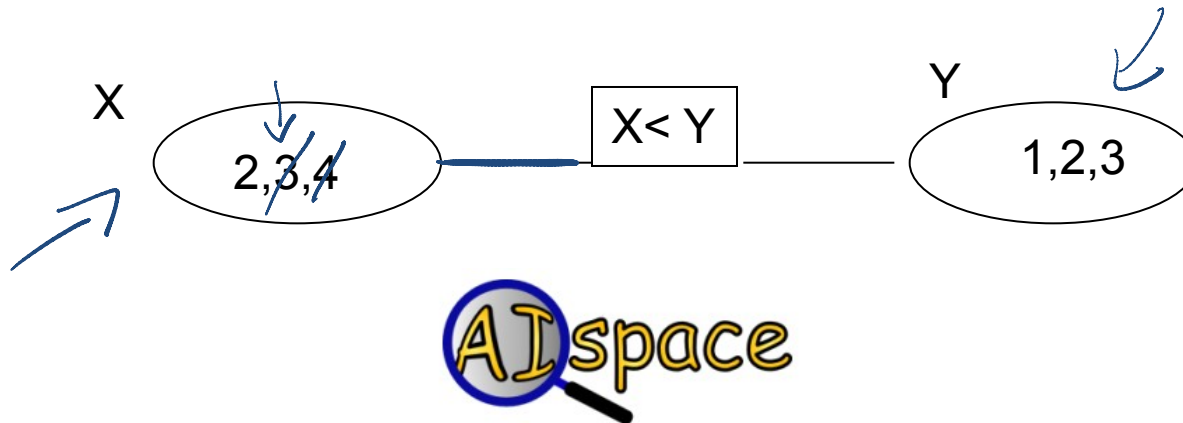
Definition (arc consistency)

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.



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 

will be posted

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