

# CSPs: Search and Arc Consistency

CPSC 322 – CSPs 2

Textbook §4.3 – 4.5

# Lecture Overview

- 1 Recap
- 2 Search
- 3 Consistency
- 4 Arc Consistency

# Variables

- We define the state of the world as an assignment of values to a set of **variables**
  - variable: a synonym for feature
  - we denote variables using capital letters
  - each variable  $V$  has a domain  $dom(V)$  of possible values
- Variables can be of several main kinds:
  - **Boolean**:  $|dom(V)| = 2$
  - **Finite**: the domain contains a finite number of values
  - **Infinite but Discrete**: the domain is countably infinite
  - **Continuous**: e.g., real numbers between 0 and 1
- We'll call the set of states that are induced by a set of variables the set of **possible worlds**

# Constraints

Constraints are restrictions on the values that one or more variables can take

- **Unary constraint:** restriction involving a single variable
  - of course, we could also achieve the same thing by using a smaller domain in the first place
- **$k$ -ary constraint:** restriction involving the domains of  $k$  different variables
  - it turns out that  $k$ -ary constraints can always be represented as binary constraints, so we'll often talk about this case
- Constraints can be specified by
  - giving a list of valid domain values for each variable participating in the constraint
  - giving a function that returns true when given values for each variable which satisfy the constraint
- A possible world **satisfies** a set of constraints if the set of variables involved in each constraint take values that are consistent with that constraint

# Constraint Satisfaction Problems: Definition

## Definition

A **constraint satisfaction problem** consists of:

- a set of variables
- a domain for each variable
- a set of constraints

## Definition

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

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# CSPs as Search Problems

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

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



# CSPs as Search Problems

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# CSPs as Search Problems

- What **search strategy** will work well for a CSP?
  - there are no costs, so there's no role for a heuristic function
  - the tree is always finite and has no cycles, so DFS is better than BFS
- How can we **prune** the DFS search tree?
  - once we reach a node that violates one or more constraints, we know that a solution cannot exist below that point
  - thus we should backtrack rather than continuing to search
  - this can yield us exponential savings over unpruned DFS, though it's still exponential

# Example

Problem:

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



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# Consistency Algorithms

- **Idea:** prune the domains as much as possible before selecting values from them.

## Definition

A variable is **domain consistent** if no value of the domain of the node is ruled impossible by any of the constraints.

- **Example:**  $dom(B) = \{1, 2, 3, 4\}$  isn't domain consistent if we have the constraint  $B \neq 3$ .



# Constraint Networks

- Domain consistency only talked about constraints involving a single variable
  - what can we say about constraints involving multiple variables?

## Definition

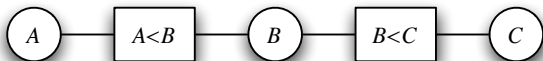
A **constraint network** is defined by a graph, with

- 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.

- When all of the constraints are binary, constraint nodes are not necessary: we can drop constraint nodes and use edges to indicate that a constraint holds between a pair of variables.
  - why can't we do the same with general  $k$ -ary constraints?

# Example Constraint Network



Recall:

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

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# Arc Consistency

## Definition

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

- In symbols,  $\forall X \in dom(X), \exists \bar{Y} \in dom(\bar{Y})$  such that  $r(X, \bar{Y})$  is satisfied.
- A network is arc consistent if all its arcs are arc consistent.
- If an arc  $\langle X, \bar{Y} \rangle$  is *not* arc consistent, all values of  $X$  in  $dom(X)$  for which there is no corresponding value in  $dom(\bar{Y})$  **may be deleted** from  $dom(X)$  to make the arc  $\langle X, \bar{Y} \rangle$  consistent.
  - This removal **can never rule out any models** (do you see why?)

# Arc Consistency Algorithm

- Consider the arcs in turn making each arc consistent.
  - An arc  $\langle X, r(X, \bar{Y}) \rangle$  needs to be revisited if the domain of  $Y$  is reduced.
- Regardless of the order in which arcs are considered, we will terminate with the same result: an arc consistent network.
- Worst-case complexity of this procedure:
  - let the max size of a variable domain be  $d$
  - let the number of constraints be  $e$
  - complexity is  $O(ed^3)$
- Some special cases are faster
  - e.g., if the constraint graph is a tree, arc consistency is  $O(ed)$

# Arc Consistency Algorithm

```

procedure AC( $V, dom, R$ )
  Inputs
     $V$ : a set of variables
     $dom$ : a function such that  $dom(X)$  is the domain of variable  $X$ 
     $R$ : set of relations to be satisfied

  Output
    arc consistent domains for each variable

  Local
     $D_X$  is a set of values for each variable  $X$ 

  for each variable  $X$  do
     $D_X \leftarrow dom(X)$ 
  end for each
   $TDA \leftarrow \{\langle X, r \rangle \mid r \in R \text{ is a constraint that involves } X\}$ 
  while  $TDA \neq \{\}$  do
    select  $\langle X, r \rangle \in TDA$ ;
     $TDA \leftarrow TDA - \{\langle X, r \rangle\}$ ;
     $ND_X \leftarrow \{x \mid x \in D_X \text{ and there is } \bar{y} \in D_{\bar{Y}} \text{ such that } r(x, \bar{y})\}$ ;
    if  $ND_X \neq D_X$  then
       $TDA \leftarrow TDA \cup \{\langle Z, r' \rangle \mid r' \neq r \text{ and } r' \text{ involves } X \text{ and } Z \neq X\}$ ;
       $D_X \leftarrow ND_X$ ;
    end if
  end while
  return  $\{D_X : X \text{ is a variable}\}$ 
end procedure

```

# Adding edges back to $TDA$

- When we change the domain of a variable  $X$  in the course of making an arc  $\langle X, r \rangle$  arc consistent, we add every arc  $\langle Z, r' \rangle$  where  $r'$  involves  $X$  and:
  - $r \neq r'$
  - $Z \neq X$
- Thus we don't add back the same arc:
  - This makes sense—it's definitely arc consistent.

# Adding edges back to $TDA$

- When we change the domain of a variable  $X$  in the course of making an arc  $\langle X, r \rangle$  arc consistent, we add every arc  $\langle Z, r' \rangle$  where  $r'$  involves  $X$  and:
  - $r \neq r'$
  - $Z \neq X$
- We don't add back other arcs that involve the **same variable**  $X$ 
  - We've just *reduced* the domain of  $X$
  - If an arc  $\langle X, r \rangle$  was arc consistent before, it will still be arc consistent
    - in the “for all” we'll just check fewer values



# Adding edges back to $TDA$

- When we change the domain of a variable  $X$  in the course of making an arc  $\langle X, r \rangle$  arc consistent, we add every arc  $\langle Z, r' \rangle$  where  $r'$  involves  $X$  and:
  - $r \neq r'$
  - $Z \neq X$
- We don't add back other arcs that involve the **same constraint** and a **different variable**:
  - Imagine that such an arc—involving variable  $Y$ —had been arc consistent before, but was no longer arc consistent after  $X$ 's domain was reduced.
  - This means that some value in  $Y$ 's domain could satisfy  $r$  only when  $X$  took one of the dropped values
  - But we dropped these values precisely because there were no values of  $Y$  that allowed  $r$  to be satisfied when  $X$  takes these values—contradiction!