

# CSPs: Search and Arc Consistency

CPSC 322 Lecture 10

January 29, 2007  
Textbook §3.3 – 3.5

# Lecture Overview

- 1 Recap
- 2 CSPs
- 3 Generate-and-Test
- 4 Search
- 5 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

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# 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.

# Constraint Satisfaction Problems: Variants

We may want to solve the following problems with a CSP:

- determine whether or not a model **exists**
- **find** a model
- **find all** of the models
- **count** the number of models
- find the **best** model, given some measure of model quality
  - this is now an optimization problem
- determine whether some **property of the variables** holds in all models

# CSPs: Game Plan

It turns out that even the simplest problem of determining whether or not a model exists in a general CSP with finite domains is  $\mathcal{NP}$ -hard

- we can't hope to find an efficient algorithm.

However, we can try to:

- find algorithms that are **fast on "typical" cases**
- identify **special cases** for which algorithms are efficient (polynomial)
- find **approximation algorithms** that can find good solutions quickly, even they may offer no theoretical guarantees
- develop **parallel or distributed algorithms** so that additional hardware can be used



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# Generate-and-Test Algorithm

- The **assignment space** of a CSP is the space of possible worlds
- Algorithm:
  - **Generate** possible worlds one at a time from the assignment space
  - **Test** them to see if they violate any constraints
- This procedure is able to solve any CSP
- However, the running time is proportional to the size of the state space
  - always exponential in the number of variables
  - far too long for many CSPs

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# CSPs as Search Problems

In order to think about better ways to solve CSPs, let's map CSPs into search problems.

- **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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  - the tree is always finite and has no cycles, so DFS is better than BFS
  - DFS is one way of implementing generate-and-test

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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
  - DFS is one way of implementing generate-and-test
- 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 generate-and-test, though it's still exponential

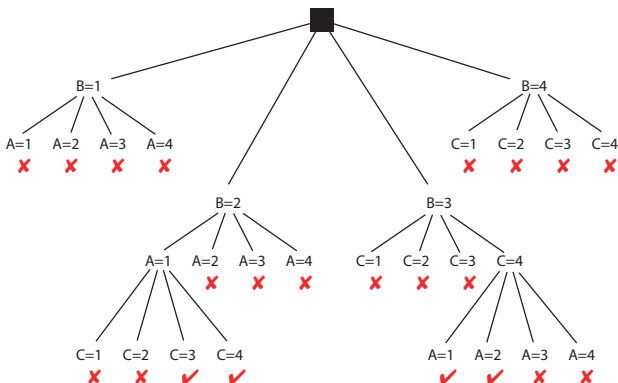


# Example

Problem:

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

# Example



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

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

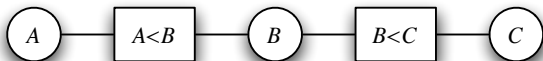
A **constraint network** is defined by a graph, with

- 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:  $A < B, B < C$