

# Constraint Satisfaction Problems

Blake VanBerlo

Lecture 4

Readings: RN 6.1 - 6.3. PM 4.1 - 4.4.

# Outline

Learning Goals

Examples of CSP Problems

Formulating a CSP

Solving a CSP

- Backtracking Search

- The Arc Consistency Definition

- The AC-3 Arc Consistency Algorithm

- Combining Backtracking and Arc Consistency

Revisiting the Learning goals

# Learning Goals

By the end of the lecture, you should be able to

- ▶ Formulate a real-world problem as a constraint satisfaction problem.
- ▶ Trace the execution of the backtracking search algorithm.
- ▶ Verify whether a constraint is arc-consistent.
- ▶ Trace the execution of the AC-3 arc consistency algorithm.
- ▶ Trace the execution of the backtracking search algorithm with arc consistency.

Learning Goals

Examples of CSP Problems

Formulating a CSP

Solving a CSP

Revisiting the Learning goals

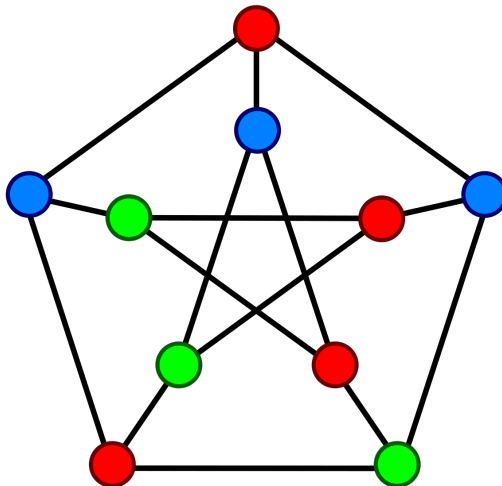
# Evacuation Planning

- ▶ Instruct residents to follow a route at a given time.
- ▶ Two challenges:
  - (1) deploy enough resources to give instructions.
  - (2) ensure that the population comply with the instructions.
- ▶ Applied in a real-life case study and generated schedules close to the optimal ones from prior work.
- ▶ Even, C., Schutt, A., & Van Hentenryck, P. (2015).  
A constraint programming approach for non-preemptive evacuation scheduling.  
<https://arxiv.org/pdf/1505.02487.pdf>.

# Crossword Puzzles



# Graph Colouring Problem

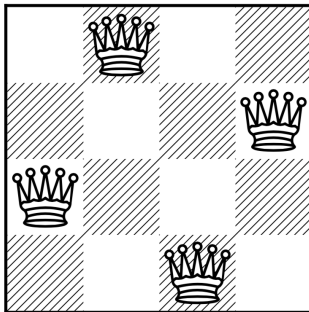


# Sudoku

5	3			7				
6			1	9	5			
	9	8					6	
8				6				3
4			8		3			1
7				2				6
	6					2	8	
			4	1	9			5
				8			7	9



# 4-Queens Problem



Learning Goals

Examples of CSP Problems

Formulating a CSP

Solving a CSP

Revisiting the Learning goals

# Internal Structure of States

- ▶ Search algorithms are unaware of the internal structure of states.

→ Generate successors of a state. Test whether a state is a goal.

- ▶ However, knowing a state's internal structure can help.

→ 4-queens: Consider a state with 2 queens in the same row. A search algorithm only knows that this is not a goal and will keep searching. But in fact, this is a dead end and we should backtrack.

Let's model the internal structure of states.

# Defining a CSP

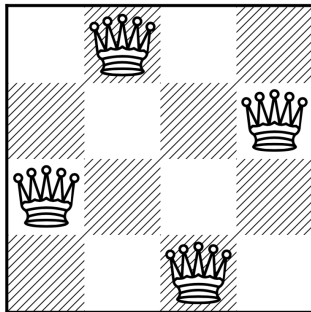
Each state contains

- ▶ A set  $X$  of variables:  $\{X_1, X_2, \dots, X_n\}$ .
- ▶ A set  $D$  of domains:  $D_i$  is the domain for variable  $X_i$ ,  $\forall i$ .
- ▶ A set  $C$  of constraints specifying allowable value combinations.

A solution is an assignment of values to all the variables that satisfy all the constraints.

→ More efficient than search because the constraints will help us eliminate large portions of the search space.

## Example: 4-Queens Problem



## 4-Queens: State Definition in a CSP

## 4-Queens: State Definition in a CSP

- Variables:  $x_0, x_1, x_2, x_3$  where  $x_i$  is the row position of the queen in column  $i$ , where  $i \in \{0, 1, 2, 3\}$ .

Assume that exactly one queen is in each column.

## 4-Queens: State Definition in a CSP

- ▶ Variables:  $x_0, x_1, x_2, x_3$  where  $x_i$  is the row position of the queen in column  $i$ , where  $i \in \{0, 1, 2, 3\}$ .

Assume that exactly one queen is in each column.

- ▶ Domains:  $D_{x_i} = \{0, 1, 2, 3\}$  for all  $x_i$ .



## 4-Queens: State Definition in a CSP

- Variables:  $x_0, x_1, x_2, x_3$  where  $x_i$  is the row position of the queen in column  $i$ , where  $i \in \{0, 1, 2, 3\}$ .

Assume that exactly one queen is in each column.

- Domains:  $D_{x_i} = \{0, 1, 2, 3\}$  for all  $x_i$ .

- Constraints:

No pair of queens are in the same row or diagonal.

$$(\forall i(\forall j((i \neq j) \rightarrow ((x_i \neq x_j) \wedge (|x_i - x_j| \neq |i - j|)))))$$

For example,  $((x_0 \neq x_1) \wedge (|x_0 - x_1| \neq 1))$

## Q: Constraints for 4-Queens Problem

**Q:** Given the definitions of variables and their domains for the 4-queens problems, which constraints do we need to define?

- (A) No two queens can be in the same row.
- (B) No two queens can be in the same column.
- (C) No two queens can be in the same diagonal.
- (D) Two of (A), (B), and (C).
- (E) All of (A), (B), and (C).

## Q: Constraints for 4-Queens Problem

**Q:** Given the definitions of variables and their domains for the 4-queens problems, which constraints do we need to define?

- (A) No two queens can be in the same row.
- (B) No two queens can be in the same column.
- (C) No two queens can be in the same diagonal.
- (D) Two of (A), (B), and (C).
- (E) All of (A), (B), and (C).

→ Correct Answer: (D) Two of A, B, and C.

No need to specify column constraint. Already defined separate variables for each column. Implicitly saying that we put one queen in each column.

# Defining Constraints

There are two ways of defining a constraint.

- ▶ The list/table format:  
Give a list/table of values of the variables that satisfy the constraints.
- ▶ The function/formula format:  
Give a function/formula, which returns/is true if the values of the variables satisfy the constraint.

→ In an implementation, the second format is a function, which returns true if the values satisfy the constraint.

## Q: Defining Constraints as a Table

**Q:** Suppose that we use a 2-column table to encode the following constraint. In each row of the table, the two values of  $x_0$  and  $x_2$  satisfy the constraint.

*The two queens in columns 0 and 2 are not in the same row or diagonal.*

How many rows are there in this table?

- (A) Less than 8
- (B) 8
- (C) 9
- (D) 10
- (E) More than 10

## Q: Defining Constraints as a Table

**Q:** Suppose that we use a 2-column table to encode the following constraint. In each row of the table, the two values of  $x_0$  and  $x_2$  satisfy the constraint.

*The two queens in columns 0 and 2 are not in the same row or diagonal.*

How many rows are there in this table?

- (A) Less than 8
- (B) 8
- (C) 9
- (D) 10
- (E) More than 10

→ Correct Answer: 8

## Q: Defining Constraints as a Formula

**Q:** How should we encode the following constraint as a propositional formula?

*The two queens in columns 0 and 2 are not in the same row or diagonal.*

(A)  $(x_0 \neq x_2)$

(B)  $((x_0 \neq x_2) \wedge ((x_0 - x_2) \neq 1))$

(C)  $((x_0 \neq x_2) \wedge ((x_0 - x_2) \neq 2))$

(D)  $((x_0 \neq x_2) \wedge (|x_0 - x_2| \neq 1))$

(E)  $((x_0 \neq x_2) \wedge (|x_0 - x_2| \neq 2))$

## Q: Defining Constraints as a Formula

**Q:** How should we encode the following constraint as a propositional formula?

*The two queens in columns 0 and 2 are not in the same row or diagonal.*

(A)  $(x_0 \neq x_2)$

(B)  $((x_0 \neq x_2) \wedge ((x_0 - x_2) \neq 1))$

(C)  $((x_0 \neq x_2) \wedge ((x_0 - x_2) \neq 2))$

(D)  $((x_0 \neq x_2) \wedge (|x_0 - x_2| \neq 1))$

(E)  $((\mathbf{x}_0 \neq \mathbf{x}_2) \wedge (|\mathbf{x}_0 - \mathbf{x}_2| \neq \mathbf{2}))$

→ Correct Answer: (E)  $((x_0 \neq x_2) \wedge (|x_0 - x_2| \neq 2))$



# Expressing Constraints

- As a propositional formula:

$$((x_0 \neq x_2) \wedge (|x_0 - x_2| \neq 2))$$

- As a table of allowable combinations of values

$x_0$	$x_1$
0	1
0	3
1	0
1	2
2	1
2	3
3	0
3	2

Learning Goals

Examples of CSP Problems

Formulating a CSP

Solving a CSP

- Backtracking Search

- The Arc Consistency Definition

- The AC-3 Arc Consistency Algorithm

- Combining Backtracking and Arc Consistency

Revisiting the Learning goals

# 4-Queens Incremental CSP Formulation

## 4-Queens Incremental CSP Formulation

- ▶ State: one queen per column in the leftmost  $k$  columns with no pair of queens attacking each other.
- ▶ Variables:  $x_0, x_1, x_2, x_3$  where  $x_i$  is the row position of the queen in column  $i$ , where  $i \in \{0, 1, 2, 3, -\}$ . Exactly one queen is in each column.  $x_i = -$  denotes that column  $i$  does not have a queen.
- ▶ Domains:  $D_{x_i} = \{0, 1, 2, 3\}$  for all  $x_i$ .
- ▶ Constraints: No pair of queens are in the same row or diagonal.

## 4-Queens Incremental CSP Formulation

- ▶ State: one queen per column in the leftmost  $k$  columns with no pair of queens attacking each other.
  - ▶ Variables:  $x_0, x_1, x_2, x_3$  where  $x_i$  is the row position of the queen in column  $i$ , where  $i \in \{0, 1, 2, 3, -\}$ . Exactly one queen is in each column.  $x_i = -$  denotes that column  $i$  does not have a queen.
  - ▶ Domains:  $D_{x_i} = \{0, 1, 2, 3\}$  for all  $x_i$ .
  - ▶ Constraints: No pair of queens are in the same row or diagonal.
- ▶ Initial state: the empty board, that is,  $- - - -$ .

## 4-Queens Incremental CSP Formulation

- ▶ State: one queen per column in the leftmost  $k$  columns with no pair of queens attacking each other.
  - ▶ Variables:  $x_0, x_1, x_2, x_3$  where  $x_i$  is the row position of the queen in column  $i$ , where  $i \in \{0, 1, 2, 3, -\}$ . Exactly one queen is in each column.  $x_i = -$  denotes that column  $i$  does not have a queen.
  - ▶ Domains:  $D_{x_i} = \{0, 1, 2, 3\}$  for all  $x_i$ .
  - ▶ Constraints: No pair of queens are in the same row or diagonal.
- ▶ Initial state: the empty board, that is,  $- - - -$ .
- ▶ Goal state: 4 queens on the board. No pair of queens are attacking each other. For example, 2 0 3 1 is a goal state.

## 4-Queens Incremental CSP Formulation

- ▶ State: one queen per column in the leftmost  $k$  columns with no pair of queens attacking each other.
  - ▶ Variables:  $x_0, x_1, x_2, x_3$  where  $x_i$  is the row position of the queen in column  $i$ , where  $i \in \{0, 1, 2, 3, -\}$ . Exactly one queen is in each column.  $x_i = -$  denotes that column  $i$  does not have a queen.
  - ▶ Domains:  $D_{x_i} = \{0, 1, 2, 3\}$  for all  $x_i$ .
  - ▶ Constraints: No pair of queens are in the same row or diagonal.
- ▶ Initial state: the empty board, that is,  $- - - -$ .
- ▶ Goal state: 4 queens on the board. No pair of queens are attacking each other. For example, 2 0 3 1 is a goal state.
- ▶ Successor function: add a queen to the leftmost empty column such that it is not attacked by any other existing queen. For example, 0  $- - -$  has two successors 0 2  $- -$  and 0 3  $- -$ .

# Backtracking Search

---

**Algorithm 1** BACKTRACK(*assignment*, *csp*)

---

- 1: **if** *assignment* is complete **then return** *assignment*
  - 2: Let *var* be an unassigned variable
  - 3: **for every** value in the domain of *var* **do**
  - 4:     **if** adding {*var* = value} satisfies every constraint **then**
  - 5:         add {*var* = value} to *assignment*
  - 6:         *result*  $\leftarrow$  BACKTRACK(*assignment*, *csp*)
  - 7:         **if** *result*  $\neq$  failure **then return** *result*
  - 8:     remove {*var* = value} from *assignment* if it was added
  - 9: **return** failure
-



# Solve 4-Queens using Backtracking Search

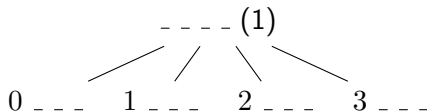
Step 0:

- - - -

	$x_0$	$x_1$	$x_2$	$x_3$
0				
1				
2				
3				

# Solve 4-Queens using Backtracking Search

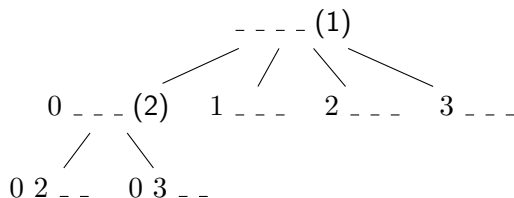
Step 1:



	$x_0$	$x_1$	$x_2$	$x_3$
0				
1				
2				
3				

# Solve 4-Queens using Backtracking Search

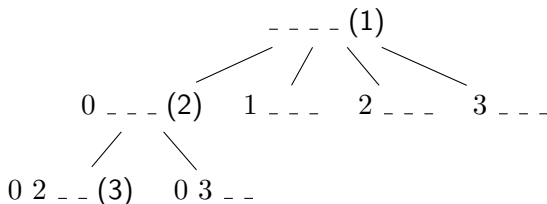
Step 2:



	$x_0$	$x_1$	$x_2$	$x_3$
0	Q	X	X	X
1	X	X		
2	X		X	
3	X			X

# Solve 4-Queens using Backtracking Search

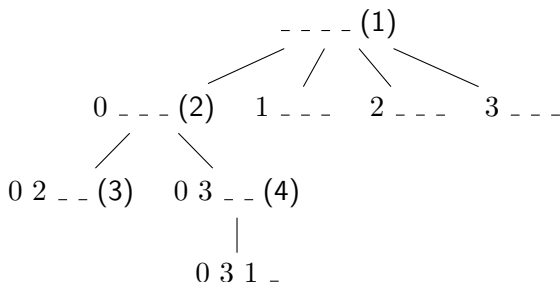
Step 3:



	$x_0$	$x_1$	$x_2$	$x_3$
0	Q	X	X	X
1	X	X	X	
2	X	Q	X	
3	X	X	X	X

# Solve 4-Queens using Backtracking Search

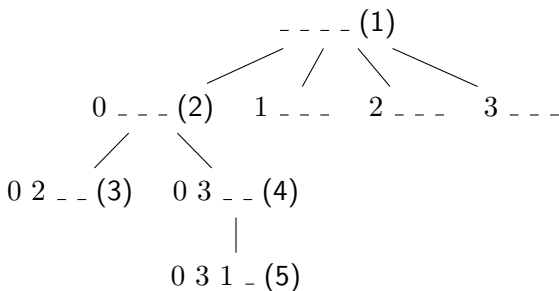
Step 4:



	$x_0$	$x_1$	$x_2$	$x_3$
0	Q	X	X	X
1	X	X		X
2	X	X	X	
3	X	Q	X	X

# Solve 4-Queens using Backtracking Search

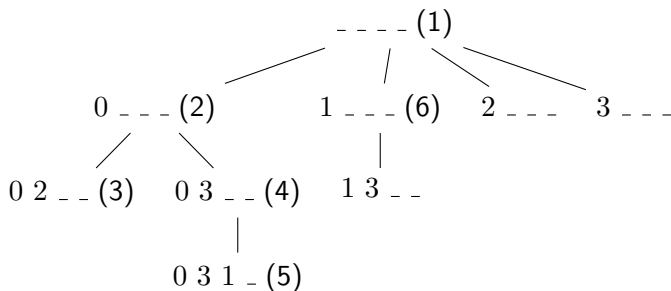
Step 5:



	$x_0$	$x_1$	$x_2$	$x_3$
0	Q	X	X	X
1	X	X	Q	X
2	X	X	X	X
3	X	Q	X	X

# Solve 4-Queens using Backtracking Search

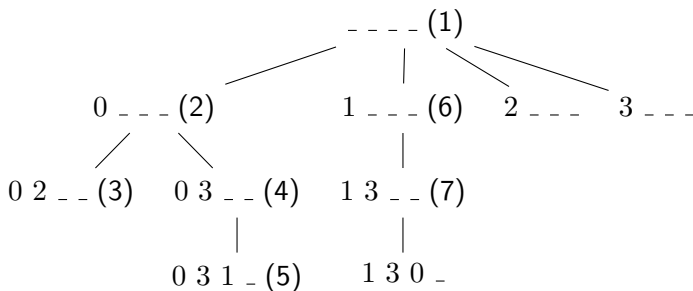
Step 6:



	$x_0$	$x_1$	$x_2$	$x_3$
0	X	X		
1	Q	X	X	X
2	X	X		
3	X		X	

# Solve 4-Queens using Backtracking Search

Step 7:

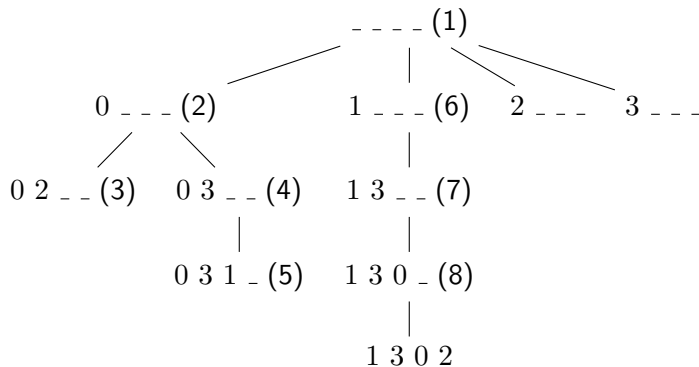


	$x_0$	$x_1$	$x_2$	$x_3$
0	X	X		
1	Q	X	X	X
2	X	X	X	
3	X	Q	X	X



# Solve 4-Queens using Backtracking Search

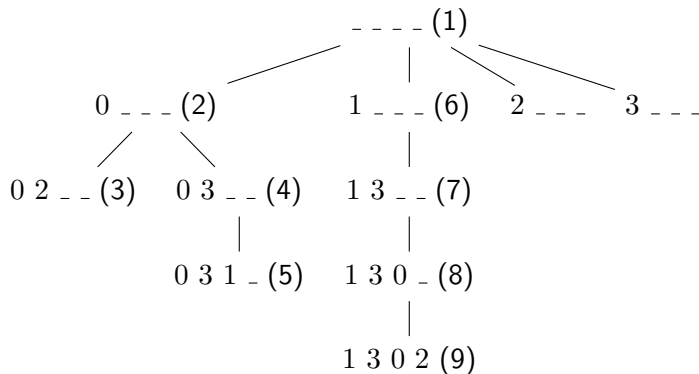
Step 8:



	$x_0$	$x_1$	$x_2$	$x_3$
0	X	X	Q	X
1	Q	X	X	X
2	X	X	X	
3	X	Q	X	X

# Solve 4-Queens using Backtracking Search

Step 9:



	$x_0$	$x_1$	$x_2$	$x_3$
0	X	X	Q	X
1	Q	X	X	X
2	X	X	X	Q
3	X	Q	X	X

Learning Goals

Examples of CSP Problems

Formulating a CSP

Solving a CSP

- Backtracking Search

- The Arc Consistency Definition

- The AC-3 Arc Consistency Algorithm

- Combining Backtracking and Arc Consistency

Revisiting the Learning goals

# The Idea of Arc Consistency

$x_0 = 0$  and  $x_1 = 2$  do not lead to a solution. Why?

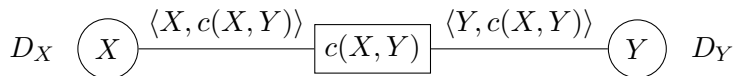
	$x_0$	$x_1$	$x_2$	$x_3$
0	Q	X	X	X
1	X	X	X	
2	X	Q	X	X
3	X	X	X	X

# Handling Different Types of Constraints

- ▶ Consider binary constraints only.
- ▶ How should we handle unary constraints?
  - Remove invalid values from variable domain
- ▶ How should be handle constraints involving 3 or more variables?
  - Convert them to binary constraints

# Notation for an Arc

- ▶  $X$  and  $Y$  are two variables.  $c(X, Y)$  is a binary constraint.



- ▶  $\langle X, c(X, Y) \rangle$  denotes an arc.  
 $X$  is the *primary variable* and  $Y$  is the *secondary variable*.

# The Arc Consistency Definition

## Definition (Arc Consistency)

An arc  $\langle X, c(X, Y) \rangle$  is arc-consistent if and only if for every value  $v \in D_X$ , there is a value  $w \in D_Y$  such that  $(v, w)$  satisfies the constraint  $c(X, Y)$ .

## Q: Applying The Arc Consistency Definition

**Q:** Consider the constraint  $X < Y$ . Let  $D_X = \{1, 2\}$  and  $D_Y = \{1, 2\}$ . Is the arc  $\langle X, X < Y \rangle$  consistent?

- (A) Yes.
- (B) No.
- (C) Maybe?
- (D) I don't know.



## Q: Applying The Arc Consistency Definition

**Q:** Consider the constraint  $X < Y$ . Let  $D_X = \{1, 2\}$  and  $D_Y = \{1, 2\}$ . Is the arc  $\langle X, X < Y \rangle$  consistent?

(A) Yes.

(B) **No.**

(C) Maybe?

(D) I don't know.

→ Correct answer is (B) No. We can remove 2 from  $D_X$ .

Learning Goals

Examples of CSP Problems

Formulating a CSP

Solving a CSP

- Backtracking Search

- The Arc Consistency Definition

- The AC-3 Arc Consistency Algorithm

- Combining Backtracking and Arc Consistency

Revisiting the Learning goals

# The AC-3 Arc Consistency Algorithm

---

**Algorithm 2** The AC-3 Algorithm

---

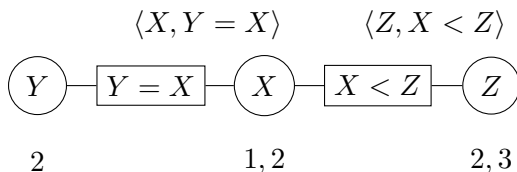
- 1: put every arc in the set  $S$ .
  - 2: **while**  $S$  is not empty **do**
  - 3:     select and remove  $\langle X, c(X, Y) \rangle$  from  $S$
  - 4:     remove every value in  $D_X$  that doesn't have a value in  $D_Y$   
      that satisfies the constraint  $c(X, Y)$
  - 5:     **if**  $D_X$  was reduced **then**
  - 6:         **if**  $D_X$  is empty **then return** false
  - 7:         for every  $Z \neq Y$ , add  $\langle Z, c'(Z, X) \rangle$  to  $S$
  - return** true
-

## Why do we need to add arcs back to $S$ ?

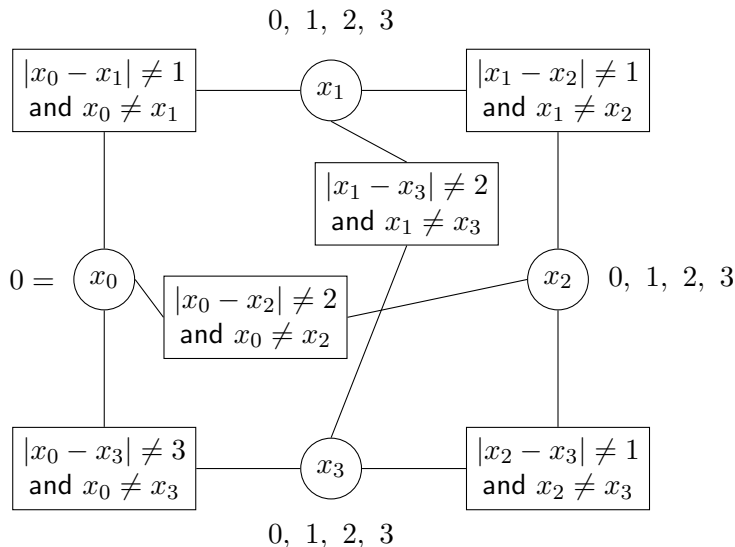
Q: After reducing a variable's domain, we may need to add arcs back to  $S$ . Why?

A: Reducing a variable's domain may cause a previously consistent arc to become inconsistent.

Example:



# Trace the AC-3 Algorithm on 4-Queens Problem



# Properties of the AC-3 Algorithm

- ▶ Does the order of removing arcs matter?

# Properties of the AC-3 Algorithm

- ▶ Does the order of removing arcs matter?  
→ No.

# Properties of the AC-3 Algorithm

- ▶ Does the order of removing arcs matter?  
→ No.
- ▶ Three possible outcomes of the arc consistency algorithm:



# Properties of the AC-3 Algorithm

- ▶ Does the order of removing arcs matter?

→ No.

- ▶ Three possible outcomes of the arc consistency algorithm:

→

1. A domain is empty. No solution.
2. Every domain has 1 value left. Found the solution without search.
3. Every domain has at least 1 value left and some domain has multiple values left. Need search to find a solution.

# Properties of the AC-3 Algorithm

- ▶ Is AC-3 guaranteed to terminate?  
→ Yes.

# Properties of the AC-3 Algorithm

- ▶ Is AC-3 guaranteed to terminate?  
→ Yes.
- ▶ What is the complexity of AC-3?

# Properties of the AC-3 Algorithm

- ▶ Is AC-3 guaranteed to terminate?

→ Yes.

- ▶ What is the complexity of AC-3? → Time complexity:

$O(cd^3)$ .

$n$  variables,  $c$  binary constraints, and the size of each domain is at most  $d$ . Each arc  $(X_k, X_i)$  can be added to the queue at most  $d$  times because we can delete at most  $d$  values from  $X_i$ . Checking consistency of each arc can be done in  $O(d^2)$  time.

Learning Goals

Examples of CSP Problems

Formulating a CSP

Solving a CSP

- Backtracking Search

- The Arc Consistency Definition

- The AC-3 Arc Consistency Algorithm

- Combining Backtracking and Arc Consistency

Revisiting the Learning goals

# Combining Backtracking and Arc Consistency

1. Perform backtracking search.
2. After each value assignment, perform arc consistency.
3. If a domain is empty, terminate and return no solution.
4. If a unique solution is found, return the solution.
5. Otherwise, continue with backtracking search on the unassigned variables.

# Solving 4-Queens Problem with Backtracking and AC-3

# Revisiting the Learning Goals

By the end of the lecture, you should be able to

- ▶ Formulate a real-world problem as a constraint satisfaction problem.
- ▶ Trace the execution of the backtracking search algorithm.
- ▶ Verify whether a constraint is arc-consistent.
- ▶ Trace the execution of the AC-3 arc consistency algorithm.
- ▶ Trace the execution of the backtracking search algorithm with arc consistency.