

# Constraint Satisfaction Problem.

A solution to CSP is assignment to each variable such that each constraint is satisfied.

Ex: N-Queens Problem

Given  $R = \langle X, D, C \rangle$

For a binary CSP  $Sol(R)$  (scope, select)

An assignment  $\bar{a}$  on scope  $S$  satisfies a constraint  $C_i$ , iff  $S_i \subseteq S$  and  $\{x_1, x_2, \dots\}$

$C_i$ , iff  $S_i \subseteq S$  and.

$$\bar{a} = \prod_{S_i} \bar{a}_{S_i} \in R_i$$

$$R_{12} = \{Q_1, Q_2\} \quad Q_3, Q_4$$

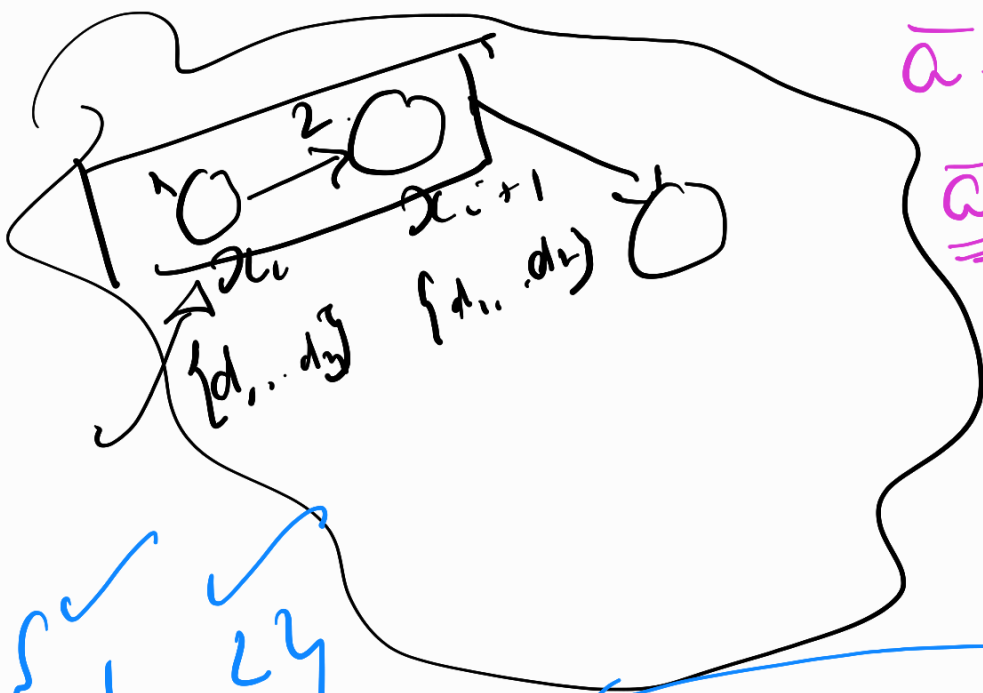
$$R_{12} = \{(3, 1)\}$$

$$R \Rightarrow \{3, 1, 2, 4\}$$

$$\bar{a} = \{3, 1\}$$

$$\bar{a} = \{3, 1, 2\}$$

$$R_{12} \quad S_i = \{Q_1, Q_2, Q_3\}$$



$$\{1, 2\}$$

$$\{1, 3, 2\}$$

$$\bar{a} = \text{Solution } \bar{a}$$

$$(K-L)^n$$

$$K$$

1				
2		$Q_2$		
3				
4	$Q_1$			

$Q_1, Q_2, Q_3, Q_4$

$\times \rightarrow$

4-Queens problem.

$$R_{12} = \{(1,3), (1,4), (3,1), (4,1), (4,2), (2,4)\}$$

$R_{23}, R_{24}, R_{13}, R_{14}, R_{34}$

$C < \underline{\text{Scope}}, \text{Rela} >$

$$\{ S: \langle (Q_1, Q_2) \leq S, R_{12} \rangle$$

$$\langle (Q_3, Q_4) \leq S, R_{34} \rangle$$

$(Q_1, Q_3)$

$$R_{13} = \{ \langle 1, 4 \rangle \}$$

$$R_{23} = \{ \langle 1, 4 \rangle \}$$

	$Q_2$		
			$Q_3$
$Q_1$			
		$Q_4$	

$\rightarrow \bar{a} = \{1, 3, 2\} \rightarrow \text{Partial Solution}$

$\rightarrow a = \{3, 1, 2, 4\}$  satisfies constraint with  $S_i$

$$R_{34} = \{ \emptyset, \emptyset, \emptyset, \emptyset \} \quad \text{4-Queens problem}$$

$$R \{x, D, c\} \rightarrow \text{Sol}(R)$$

$$\underline{R \{x, D, c\}} = \left[ \begin{array}{l} (R_{12}), (R_{13}), \\ (R_{14}), (R_{23}), \\ (R_{24}), (R_{34}) \end{array} \right]$$

$\text{Sol}(R) \downarrow$   
 $\frac{a}{a} \in$   
 $R_i \rightarrow \{a\}$

$$R_{12} = \{ \overset{\checkmark}{R_1}, \overset{\checkmark}{R_2} \}$$

$$R_{23} = \left\{ \frac{R_{12}}{R_1, R_2}, \frac{R_{13}}{R_1, R_3}, \frac{R_{23}}{R_2, R_3} \right\}$$

Backtracking ( $x, D, c$ ):

$$i \leftarrow 1$$

$$D_i' \leftarrow D_i$$

$$a = ( )$$

While  $1 \leq i \leq n$ .

$x$   $\leftarrow$  select  $(x_i, D_i', c)$ .

If  $x == \text{NULL}$ .

$i \leftarrow i - 1$

Else.

$i \leftarrow i + 1$

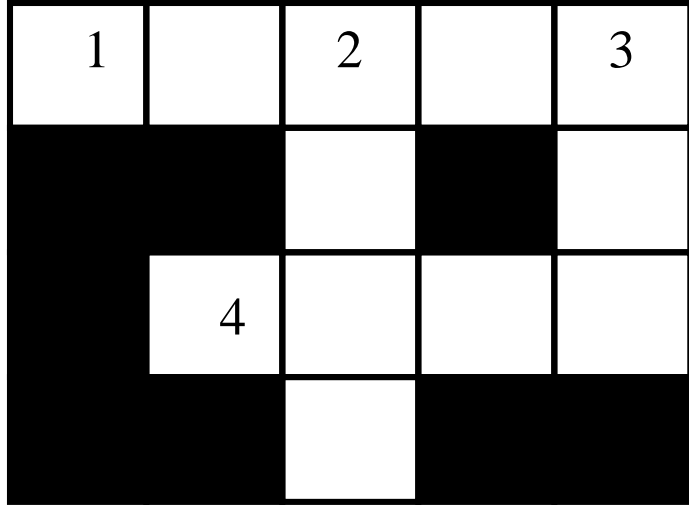
$D_i' \leftarrow D_i$

$a = (a, x)$ ,

# CSP Solution Overview

- CSP Graph Creation:
  - Create a Node for Every Variable. All possible Domain Values are initially Assigned to the Variable
  - Draw edges between Nodes if there is a Binary Constraint. Otherwise Draw a hyper-edge between nodes with constraints involving more than two variables
- Constraint Propagation:
  - Reduce the Valid Domains of Each Variable by Applying Node Consistency, Arc / Edge Consistency, K-Consistency, till no further reduction is possible. If a solution is found or the problem found to have no consistent solution, then terminate
- Search for Solution:
  - Apply Search Algorithms to Find Solutions
  - There are interesting properties of CSP graphs which lead of efficient algorithms in some cases: Trees, Perfect Graphs, Interval Graphs, etc
  - Issues for Search: Backtracking Scheme, Ordering of Children, Forward Checking (Look-Ahead) using Dynamic Constraint Propagation
  - Solving by Converting to Satisfiability (SAT) problems

# CSP Graph for Crossword



## Word List:

astar, happy, hello,  
hoses, live, load, loom,  
peal, peel, save, talk,  
ant, oak, old

# Constraint Propagation Steps

- Constraints
  - Unary Constraints or Node Constraints
  - Binary Constraints or Edges between CSP Nodes
  - Higher order or Hyper-Edges between CSP Nodes
- Node Consistency
  - For every Variable  $V_i$ , remove all elements of  $D_i$  that do not satisfy the Unary Constraints for the Variable
  - First Step is to reduce the domains using Node Consistency
- Arc Consistency
  - For every element  $x_{ij}$  of  $D_i$ , for every edge from  $V_i$  to  $V_j$ , remove  $x_{ij}$  if it has no consistent value(s) in other domains satisfying the Constraints
  - Continue to iterate using Arc Consistency till no further reduction happens.
- K-Consistency or Path Consistency
  - For every element  $y_{ij}$  of  $D_i$ , choose a Path of length  $L$  with  $L$  variables, use a consistency checking method similar to above to reduce domains if possible

# CSP Graph for Crossword

1		2		3
	4			

## Word List:

astar, happy, hello,  
hoses, live, load, loom,  
peal, peel, save, talk,  
ant, oak, old

## Applying Node Consistency:

D1 = {astar, happy, hello, hoses}

D2 = {live, load, loom, peal, peel, save, talk}

D3 = {ant, oak, old}

D4 = {live, load, loom, peal, peel, save, talk}

## NOW APPLY ARC CONSISTENCY

## Applying Arc Consistency:

D1 = {astar, happy, hello, hoses}

D2 = {live, load, loom, peal, peel, save, talk}

D3 = {ant, oak, old}

D4 = {live, load, loom, peal, peel, save, talk}



# Arc Consistency Algorithm AC-3

AC-3(*csp*) // inputs - CSP with variables, domains, constraints

1. *queue*  $\leftarrow$  local variable initialized to all arcs in *csp*
2. **while** *queue* is not empty **do**
3.      $(X_i, X_j) \leftarrow \text{pop}(\text{queue})$
4.     **if** Revise(*csp*,  $X_i, X_j$ ) **then**
5.         **if** size of  $D_i = 0$  **then return false**
6.         **for each**  $X_k$  **in**  $X_i.\text{neighbors}-\{X_j\}$  **do**
7.             add  $(X_k, X_i)$  to *queue*
8.     **return true**

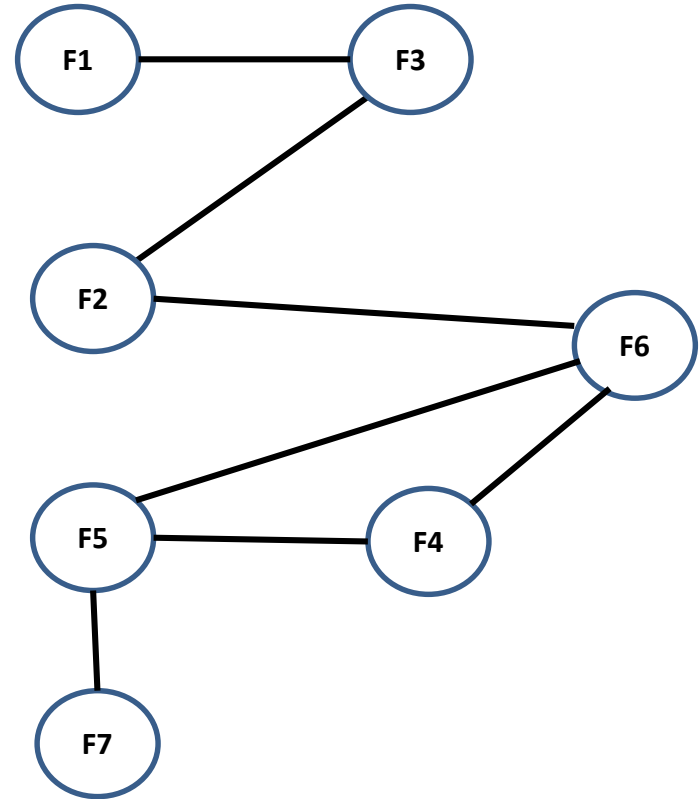
Revise(*csp*,  $X_i, X_j$ )

1. *revised*  $\leftarrow$  false
2. **for each**  $x$  **in**  $D_i$  **do**
3.     **if** no value  $y$  in  $D_j$  allows  $(x, y)$  to satisfy constraint between  $X_i$  and  $X_j$  **then**
4.         delete  $x$  from  $D_i$
5.     *revised*  $\leftarrow$  true
6. **return revised**

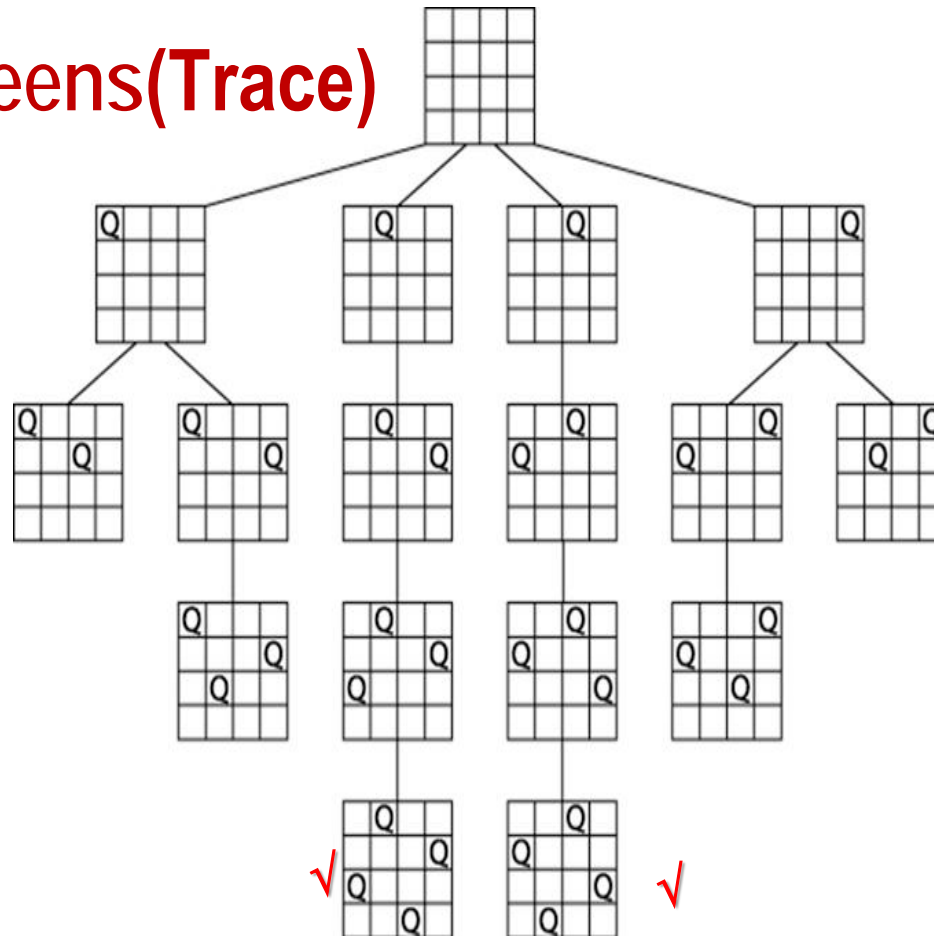
Time complexity:  $O(cd^3)$

# Backtracking for Airline Gate Scheduling

Flight No	Dep Time	G Start	G End
F1	7:00	6:15	7:15
F2	8:30	7:45	8:45
F3	7:45	7:00	8:00
F4	9:45	9:00	10:00
F5	10:00	9:15	10:15
F6	9:00	8:15	9:15
F7	11:00	10:15	11:15



# Search 4-Queens(Trace)



# Strategies for CSP Search Algorithms

- Initial Constraint Propagation
- Backtracking Search
  - Variable Ordering
    - Most Constrained Variable / Minimum Remaining Values
    - Most Constraining Variable
  - Value Ordering
    - Least Constraining Value leaving maximum flexibility
  - Dynamic Constraint Propagation Through Forward Checking
    - Preventing useless Search ahead
  - Dependency Directed Backtracking
- SAT Formulations and Solvers
- Optimization
  - Branch-and-Bound
  - SMT Solvers, Constraint Programming
- Learning, Memoizing, etc
- CSP Problems are NP-Hard in General