

# Constraint Satisfaction Problems (CSPs)

## Introduction

Computer Science cpsc322, Lecture 11

*(Textbook Chpt 4.0 – 4.2)*



January, 28, 2009

# Announcements

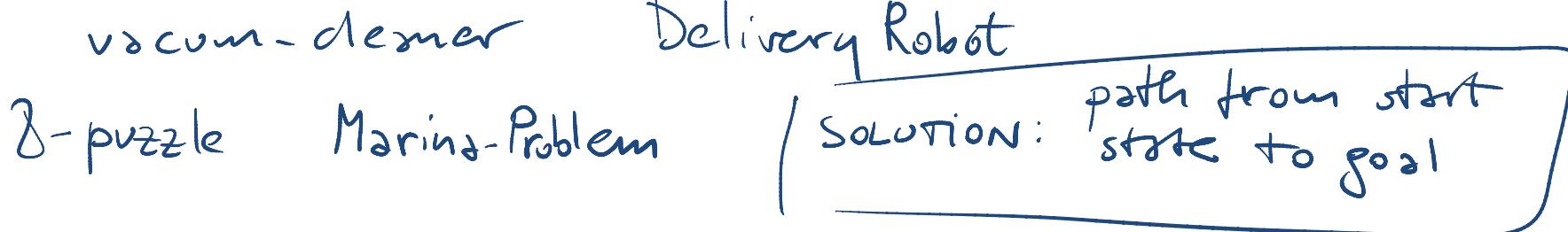
- Only one more week for assignment1
- Search wrap-up
  - Go back to **learning goals** (end of slides)
  - Make sure you understands the **inked slides**
  - More details or different examples **on textbook**
  - Work on the **practice exercises**
  - If still confused, come to **office hours**

# Lecture Overview

- **Generic Search vs. Constraint Satisfaction Problems**
- Variables
- Constraints
- CSPs

# Standard Search

To learn about search we have used it as the *reasoning strategy* for a simple goal-driven planning agent, but...



Standard search problem: An agent can solve a problem by searching in a space of states

- state is a "black box" – any arbitrary data structure that supports three problem-specific routines

goal( $u$ )      heuristic( $u$ )  
successor/neighbor( $u$ )

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

Problem	Environment	
	Deterministic	Stochastic
Static	Arc Consistency <i>Vars + Constraints</i> Search	<i>for CSP</i>
	<i>Logics</i> Search	<i>for Inference</i> Var. Elimination
Sequential	<i>STRIPS actions</i> Search	<i>for complex planning</i> Var. Elimination
		Decision Nets Markov Processes Value Iteration
Representation Reasoning Technique		

# Standard Search vs. Specific R&R systems

Constraint Satisfaction (Problems):

- State
- Successor function
- Goal test
- Solution

Planning :

- State
- Successor function
- Goal test
- Solution

Inference

- State
- Successor function
- Goal test
- Solution

} next <sup>two</sup> lectures

} following weeks

# Lecture Overview

- Generic Search vs. Constraint Satisfaction Problems
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# Variables/Features, domains and Possible Worlds

- Variables / features

- we denote variables using capital letters  $A, B$
- each variable  $V$  has a **domain**  $\text{dom}(V)$  of possible values

$$\text{dom}(A) = \text{dom}(B) = \{0, 1\}$$

- Variables can be of several main kinds:

- Boolean:  $|\text{dom}(V)| = 2$  called propositions
- Finite: the domain contains a finite number of values
- ~~Infinite but Discrete~~: the domain is countably infinite ~~not in this~~
- ~~Continuous~~: e.g., real numbers between 0 and 1 ~~course~~
- Possible world: a complete assignment of values to a set of variables

e.g.  $\{A = 0, B = 1\}$

# Possible Worlds

## Mars Explorer Example

- Weather  $\{S, C\}$
- Temperature  $\{-40^\circ, -30^\circ, +40^\circ\}$
- LocX  $0^\circ 35^\circ$  longitude  
LocY  $0^\circ 179^\circ$  latitude

sunny      cloudy      one possible state  $\{\underline{S}, \underline{+35}, \underline{30^\circ}, \underline{110^\circ}\}$

$$2 * [81 * 360 * 180]$$

number of possible worlds  
mutually exclusive

(i.e. number of possible values)

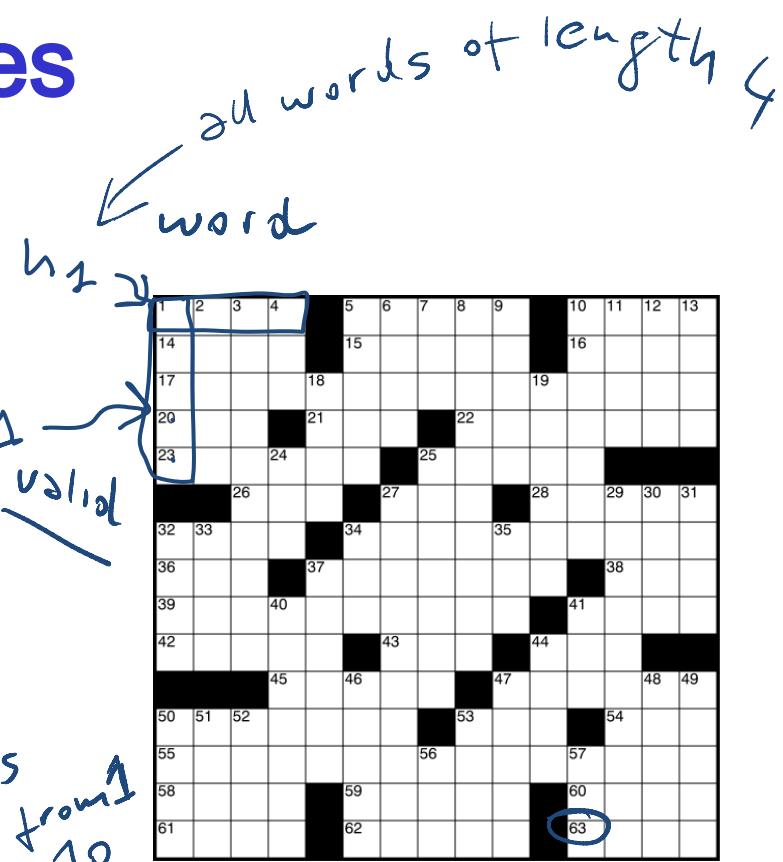
product of cardinality  
of each domain

so it is always exponential in  
the number of variables!

# Examples

- Crossword Puzzle:

- variables are words that have to be filled in
- domains are valid English words of required length
- possible worlds: all ways of assigning words



$\sim 150 * 10^3$  number of English words  
"assuming some number of words for each length from 1 to 10"  
we have  $\approx N \sim 150 * 10^2$

So # of possible words  $\approx 150 * 10^2$

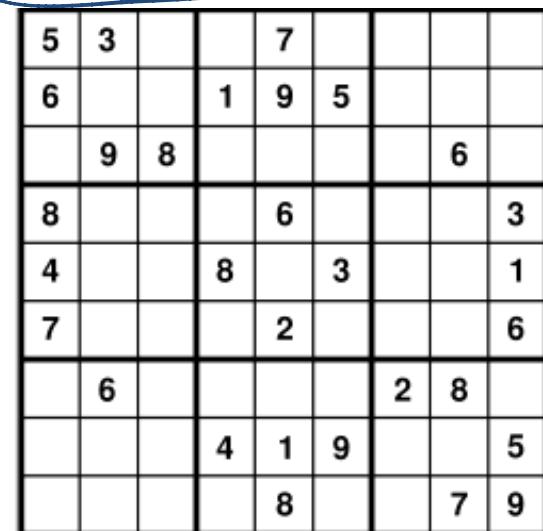
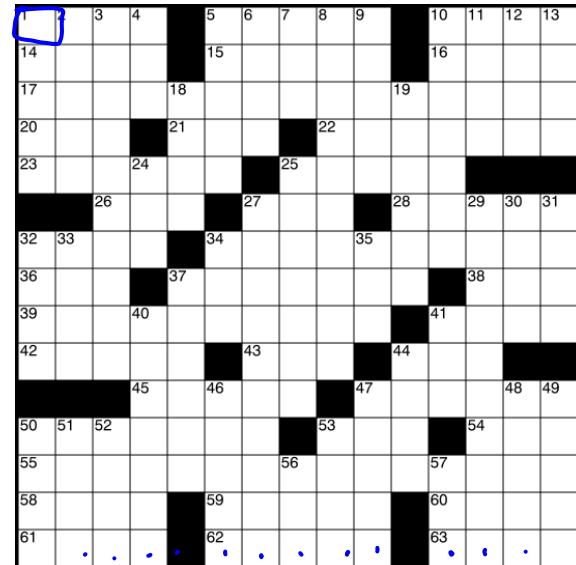
words of length 1

$(150 * 10^2)^{63}$   $\approx$  cardinality of domain of vars  
approx # of vars

# More Examples



- Crossword 2:  
• variables are cells (individual squares)  
• domains are letters of the alphabet  
• possible worlds: all ways of assigning letters to cells  
# of possible worlds  $\sim 2^6$
  - Sudoku:  
• variables are cells  
• domains are numbers between 1 and 9  
• possible worlds: all ways of assigning numbers to cells  
# of possible worlds =  $9^{\# \text{empty cells}}$



# More examples

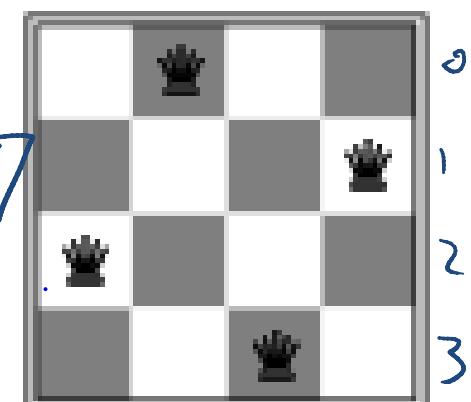
- n-Queens problem

$n$        $Q_1$      $Q_2$      $Q_n$

- variable: location of a queen on a chess board
  - there are  $n$  of them in total, hence the name
- domains: grid coordinates  $n^2$  ( $i, j$ )
- possible worlds: locations of all queens

$$\binom{n^2}{n} = \frac{n^2!}{(n^2-n)! n!}$$

$$\frac{16!}{12! 4!} \Rightarrow$$



possible ways to choose  
 $n$  location out of  $n^2$

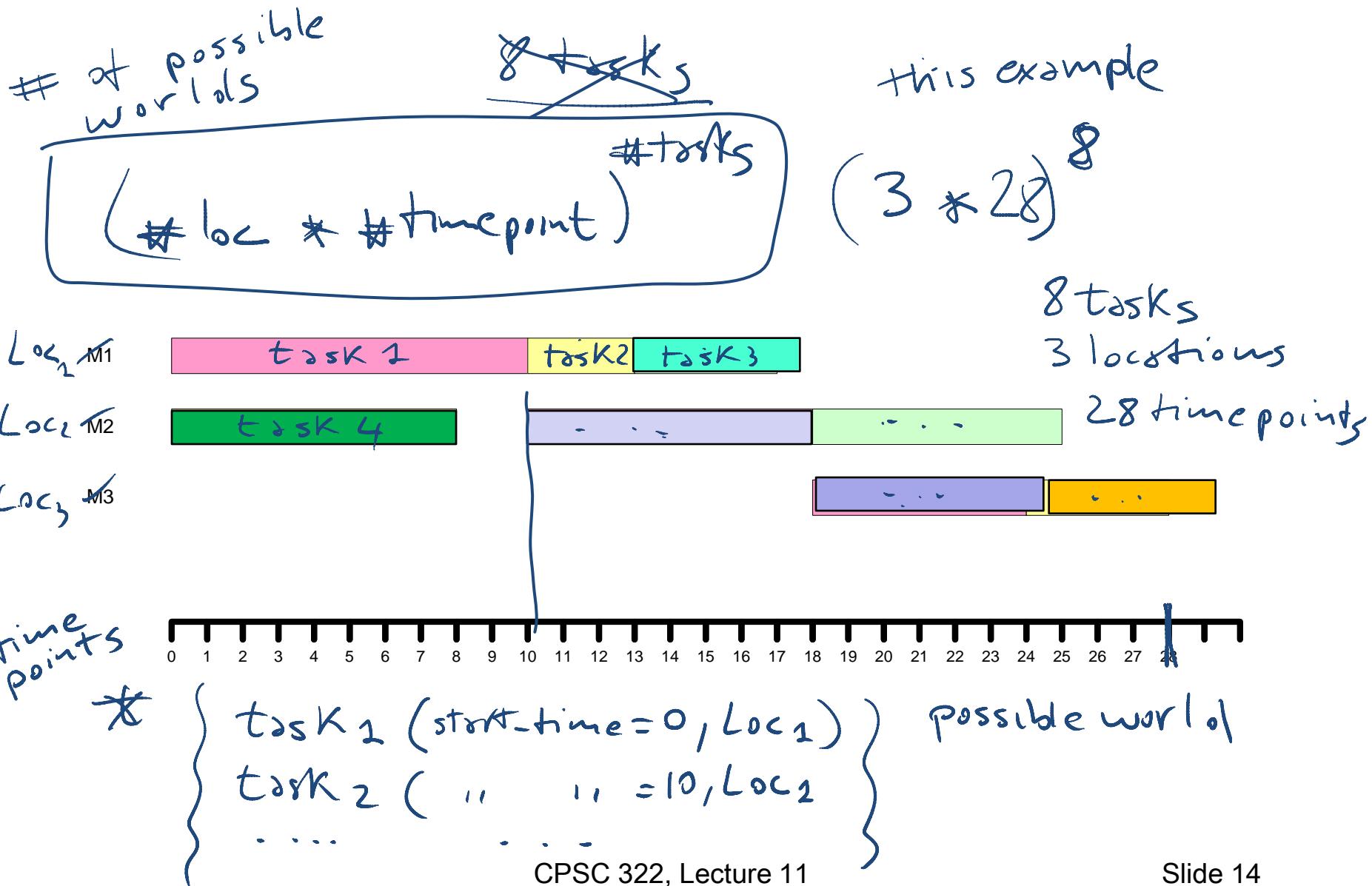
16 Possible locations:  
 $(0,0), (0,1) \dots \dots (3,3)$

# More examples

- Scheduling Problem:
  - **variables** are different tasks that need to be scheduled (e.g., course in a university; job in a machine shop)  
*assume time length of task is fixed, you only  
need to specify start-time*
  - **domains** are the different combinations of times and locations for each task (e.g., time/room for course; time/machine for job) *(start-time, location)*
  - **possible worlds:** time/location assignments for each task

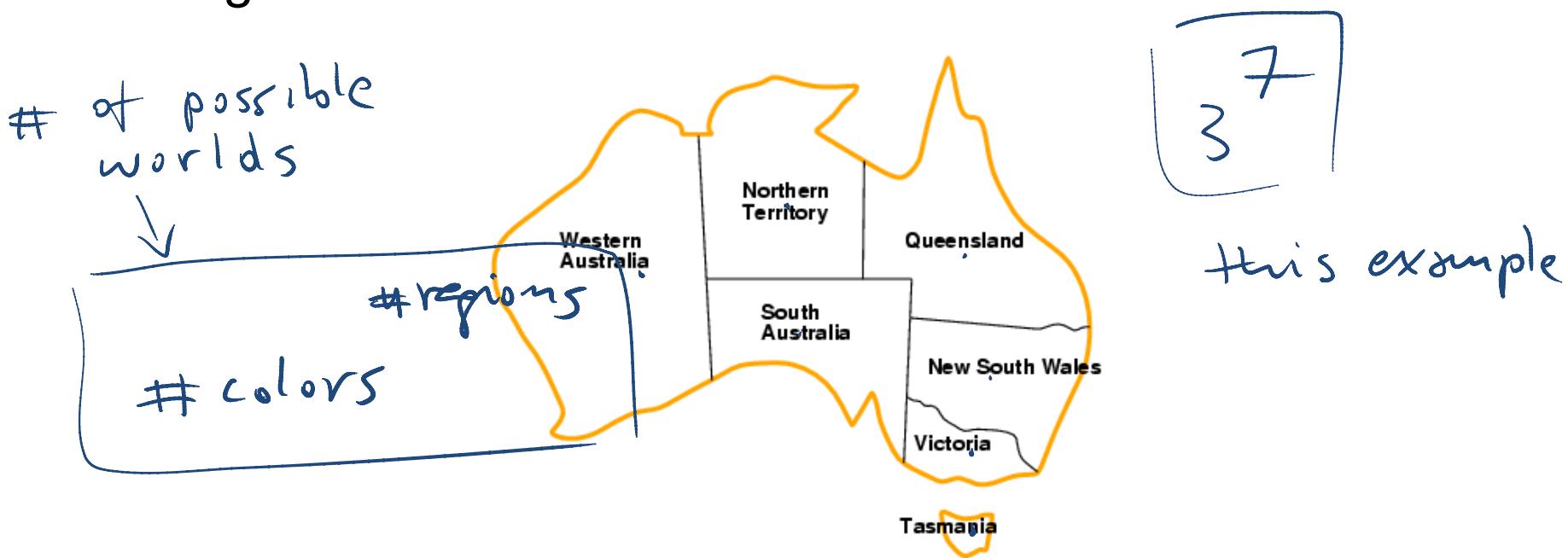
e.g.  $\text{task}_1 = \{ \text{11am...}, \text{room 310} \}$

# Scheduling possible world



# More examples....

- Map Coloring Problem
  - variable: regions on the map ~~# regions~~ 7
  - domains: possible colors ~~# colors~~ 3
  - possible worlds: color assignments for each region



# Lecture Overview

- Generic Search vs. Constraint Satisfaction Problems
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# Constraints

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

- **Unary constraint:** restriction involving a single variable

$$A = 0 \quad A, B, C \in \{0, 1\} \text{ same domain}$$

- **k-ary constraint:** restriction involving the domains of k different variables  $A > B$   $A > B + C$ 
  - it turns out that k-ary constraints can always be represented as binary constraints, so we'll *probably* only talk about this case

- **Constraints can be specified by**

- giving a list of valid domain values for each variable participating in the constraint  $\{A=0, B=1\} \{A=1, B=1\}$
- giving a function that returns true when given values for each variable which satisfy the constraint  $A = B$

# Example: Map-Coloring



Variables  $WA, NT, Q, NSW, V, SA, T$

Domains  $D_i = \{\text{red}, \text{green}, \text{blue}\}$

Constraints: adjacent regions must have different colors

e.g.,  $(WA, NT)$  in  $\{(red, green), (red, blue), (green, red), (green, blue), (blue, red), (blue, green)\}$

or  $WA \neq NT$ ,

# Constraints (cont.)

- 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

- A,B,C domains [1 .. 10]
- $A = 1, B = 2, C = 10$  *possible world*
- Constraint set1  $\{A = B, C > B\}$  *does not satisfy*
- Constraint set2  $\{A \neq B, C > B\}$  *satisfies*

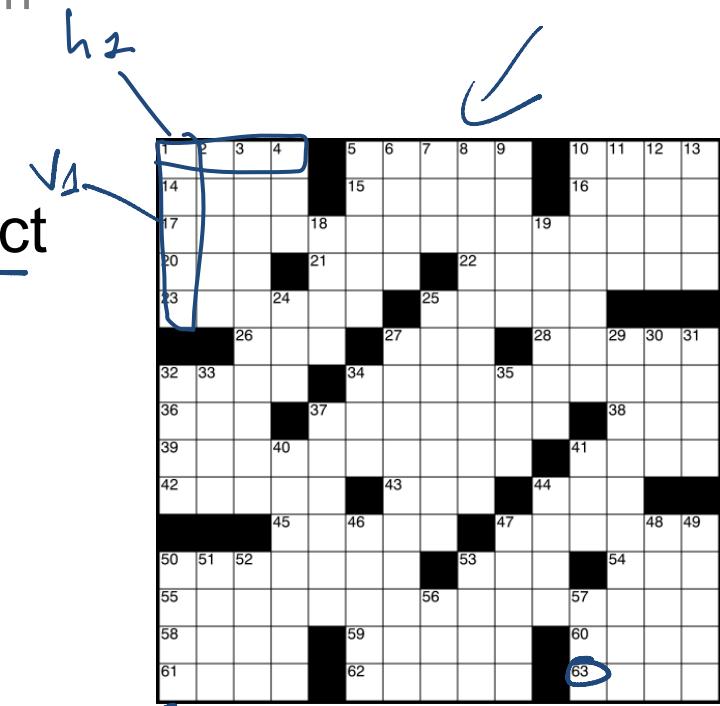
# Examples

$$h_1[\circ] = v_1[\circ]$$

- Crossword Puzzle:

- variables are words that have to be filled in
- domains are valid English words
- constraints: words have the same letters at points where they intersect

$\sim 225$  constraints



- Crossword 2:

- variables are cells (individual squares)
- domains are letters of the alphabet
- constraints: sequences of letters form valid English words

$\sim 63$  constraints  
 $\text{conc}(A[\circ, \circ], \dots, A[\circ, \circ]) \in$  4 letter English word

# Examples

- Sudoku:

- variables are cells
- domains are numbers between 1 and 9
- constraints:* rows, columns, boxes contain all different numbers

$A$

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

# of constraint =  
# empty-cells \* 24

$$A[0,2] \neq A[0,1], A[0,2] \neq A[0,0], \dots$$

solution

for all  $i$   
 $i \neq 2$

$$A[2,0] \neq A[i,0]$$

which means

$$A[2,0] \neq A[0,0]$$
$$A[2,0] \neq A[1,0]$$

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

# More examples

- n-Queens problem
  - variable: location of a queen on a chess board
    - there are  $n$  of them in total, hence the name
  - domains: grid coordinates
  - ***constraints:*** no queen can attack another
- Scheduling Problem:
  - variables are different tasks that need to be scheduled (e.g., course in a university; job in a machine shop)
  - domains are the different combinations of times and locations for each task (e.g., time/room for course; time/machine for job)
  - ***constraints:*** e.g.  $\text{task}_1(\text{loc}_1, \text{start-t}_1)$  if  $\text{start-t}_1 = \text{start-t}_2$   
 $\text{task}_2(\text{loc}_2, \text{start-t}_2)$  then  $\text{loc}_1 \neq \text{loc}_2$ 
    - ✓ tasks can't be scheduled in the same location at the same time;
    - ✓ certain tasks can be scheduled only in certain locations;
    - ✓ some tasks must come earlier than others; etc.

eg two Queens  
cannot be  
on the same  
column / row

$$Q_1 = \{x_1, y_1\}$$

$$Q_2 = \{x_2, y_2\}$$

$$x_1 = x_2 \text{ and}$$

$$y_1 = y_2$$

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# Constraint Satisfaction Problems: definitions

Definition (Constraint Satisfaction Problem)

A constraint satisfaction problem consists of

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

possible worlds

Definition (model / solution)

possible world

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

# Example: Map-Coloring



Variables  $WA, NT, Q, NSW, V, SA, T$

Domains  $D_i = \{\text{red}, \text{green}, \text{blue}\}$

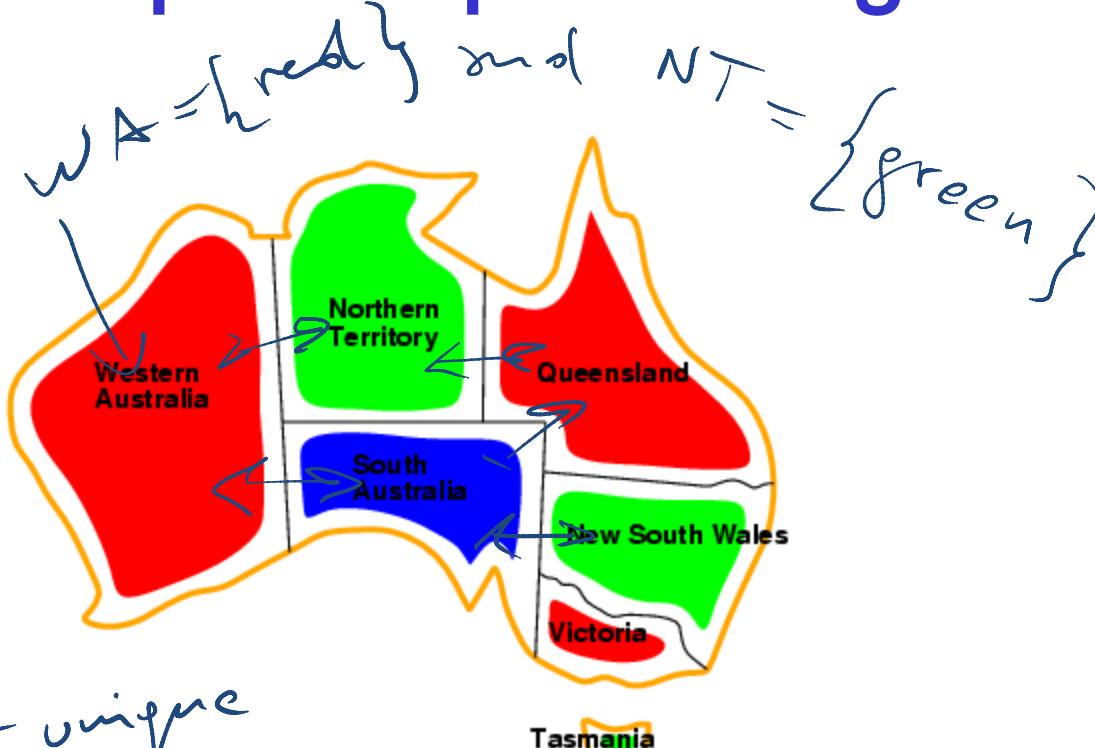
Constraints: adjacent regions must have different colors

e.g.,  $WA \neq NT$ , or

$(WA, NT) \in \{(red, green), (red, blue), (green, red), (green, blue), (blue, red), (blue, green)\}$

# Example: Map-Coloring

it becomes unique if we add two more constraints



**Models / Solutions** are complete and consistent assignments, e.g., WA = red, NT = green, Q = red, NSW = green, V = red, SA = blue, T = green

# Constraint Satisfaction Problem: Variants

We may want to solve the following problems using a CSP

- A • determine whether or not a model **exists** ← useful to avoid wasting time on B
- B • find a model our focus ←
- C • find all of the models
- D • count the number of the models
- E • find the **best** model given some model quality
  - this is now an optimization problem
- F • determine whether some **properties of the variables hold in all models**

# To summarize

- Need to think of search beyond simple goal driven planning agent.
- We started exploring the first AI Representation and Reasoning framework: CSPs



## Next class

**CSPs: Search and Arc Consistency**  
*(Textbook Chpt 4.3-4.5)*

# Learning Goals for today's class

- Define possible worlds in term of variables and their domains.
- Compute number of possible worlds on real examples
- Specify constraints to represent real world problems differentiating between:
  - Unary and k-ary constraints
  - List vs. function format.

Verify whether a possible world satisfies a set of ↗ constraints (i.e., whether it is a model, a solution)

# Extra slide (may be used here?)

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			

A possible **start state**  
(partially completed grid)

**Goal state:** 9×9 grid completely filled so that

- each column,
- each row, and
- each of the nine 3×3 boxes
- contains the digits from 1 to 9, only *one* time each

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