

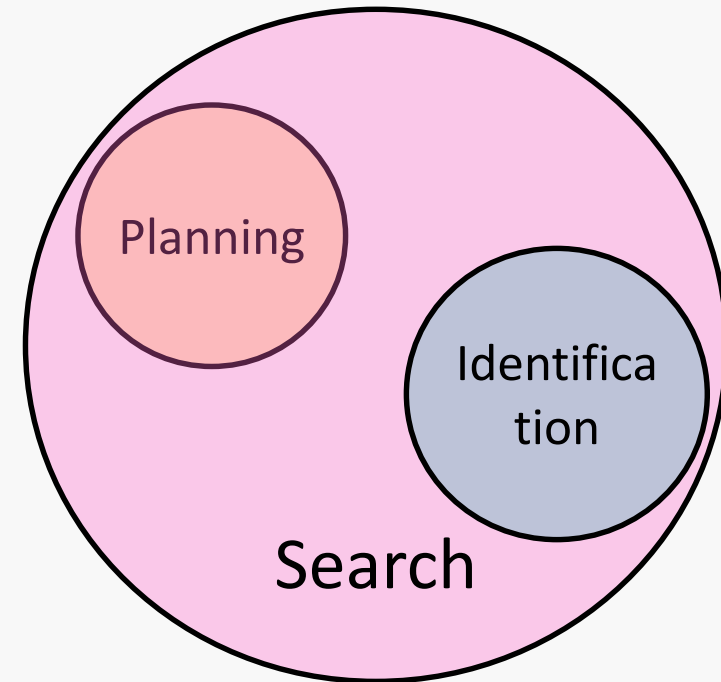


# Constraint Satisfaction Problems

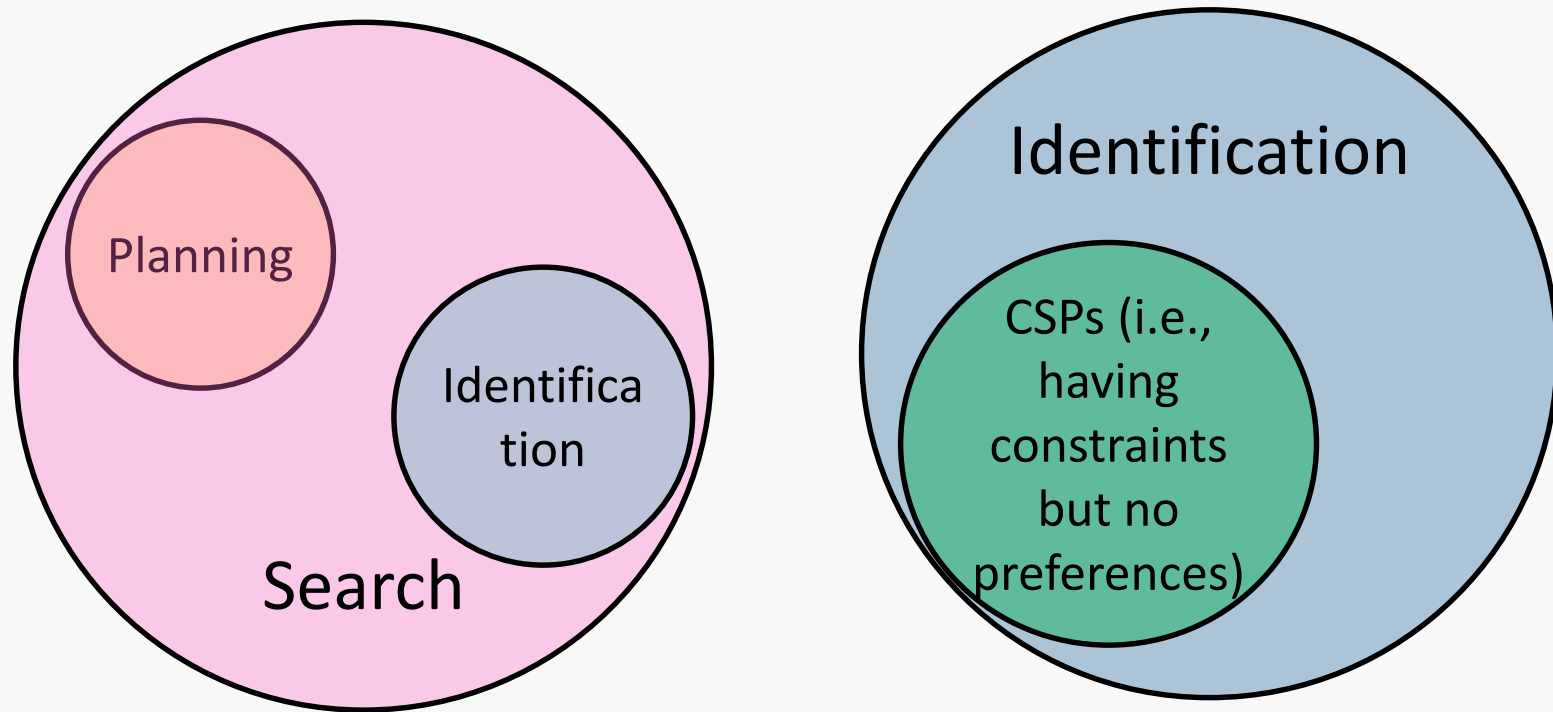


# Another type of search problems: Identification

- ◆ All the above cases are about planning, which is only one type of search problems.
- ◆ Planning: a sequence of actions.
  - ◆ We care about the path to the goal.
- ◆ Identification: an assignment
  - ◆ We care about the goal itself, not the path.
  - ◆ For example, a taxi firm assigns taxis  $a, b, c$  to customers  $x, y, z$  such that the cost incurred (e.g., fuel) is minimal.



# Search -> Identification -> CSP



- ◆ Constraint Satisfaction Problems (CSPs): Identification problems have constraints to be satisfied; there is no preference in CSPs.
- ◆ Constraints refer to hard constraints which a legal solution cannot violate.
- ◆ Preferences sometimes are referred to as soft constraints (or objectives), where we need to optimise, e.g., to minimise cost.

# CSPs

- ◆ A constraint satisfaction problem consists of
  - ◆ A set of variables
  - ◆ A domain for each variable
  - ◆ A set of constraints
- ◆ In a CSP, an assignment is **complete** if every variable has a value, otherwise it is **partial**. **Solutions** are complete assignments satisfying all the constraints.
- ◆ Example: Module scheduling problem
  - ◆ **Variables** – Modules: AI1, Data Structure, Software Engineering, OOP, AI2, Neural Computation, Evolutionary Computation...
  - ◆ **Domain** - year-term: {1-1, 1-2, 2-1, 2-2, 3-1, 3-2}
  - ◆ **Constraints** - pre-requisites: {AI1 < AI2, OOP < SE...}
  - ◆ **Solutions** - e.g.: (AI1=1-2, OOP=1-1, AI2=2-2, SE=2-1...)

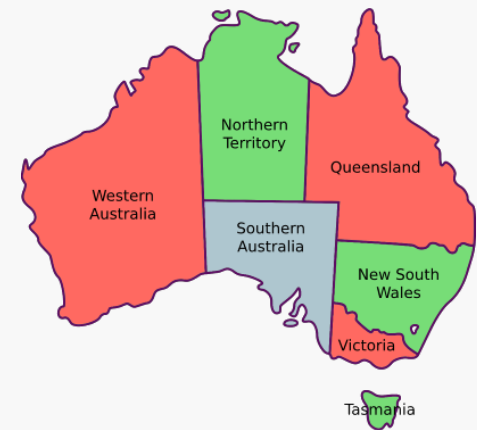
# Standard Search Problems vs CSPs

- ◆ Standard Search problems
  - ◆ State is a “black-box”: arbitrary data structure
  - ◆ Goal test can be any function over states
- ◆ Constraint Satisfaction Problems (CSPs)
  - ◆ State is defined by **variables**  $X_1, X_2, \dots$  with values from **domains**  $D_1, D_2, \dots$
  - ◆ Goal test is a set of **constraints** specifying allowable combinations of values of variables.
  - ◆ An example of a formal representation language, in which many search problems can be abstracted. This allows useful general-purpose algorithms with more power than standard search algorithms.

# Example: Map Colouring for Australia

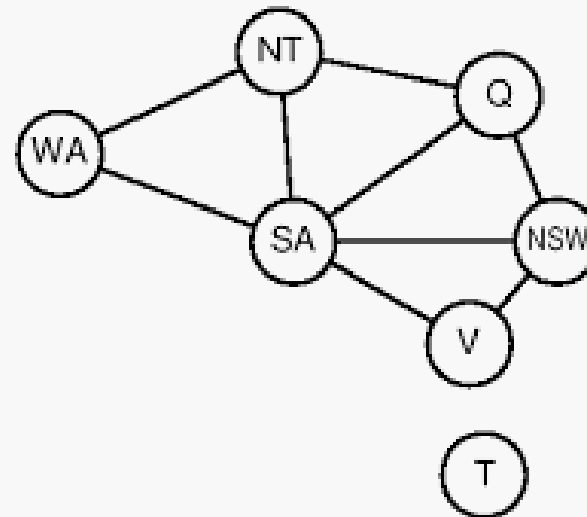
**Problem:** Map colouring problem is to paint a map (e.g., via three colours red, green and blue) in such a way that none of adjacent regions can have the same colour.

- ◆ **Variables:** WA, NT, Q, NSW, V, SA, T
- ◆ **Domain:**  $D = \{\text{red, green, blue}\}$
- ◆ **Constraints:** adjacent regions must have different colours
  - ◆  $WA \neq NT, WA \neq SA, NT \neq SA, NT \neq Q, \dots$
- ◆ **Solutions:** e.g.  $\{WA=\text{red}, NT=\text{green}, Q=\text{red}, NSW=\text{green}, V=\text{red}, SA=\text{blue}, T=\text{green}\}$



# Constraint Graphs

- ◆ Constraint graphs are used to represent relations among constraints in CSPs, where nodes correspond to the variables and arcs reflect the constraints.



What is the difference between a constraint graph and a search state graph?

# Example: Einstein Puzzle

**Problem:** There are two houses, and each house has a different colour (either blue or red), and a different pet (cat, dog or fish). We also have the following constraints: The first house does not have a dog nor fish. The blue house has a fish. Which colour and pet each house has?

- ◆ **Variables:** Colour1, Colour2, Pet1, Pet2
- ◆ **Domain:**
  - ◆ Colour1, Colour2 = {blue, red}
  - ◆ Pet1, Pet2 = {cat, dog, fish}
- ◆ **Constraints:**
  - ◆ Pet1  $\neq$  dog, fish
  - ◆ If Colour1 = blue, then Pet1 = fish
  - ◆ If Colour2 = blue, then Pet2 = fish
  - ◆ Colour1  $\neq$  Colour2, Pet1  $\neq$  Pet2



# Example: Sudoku

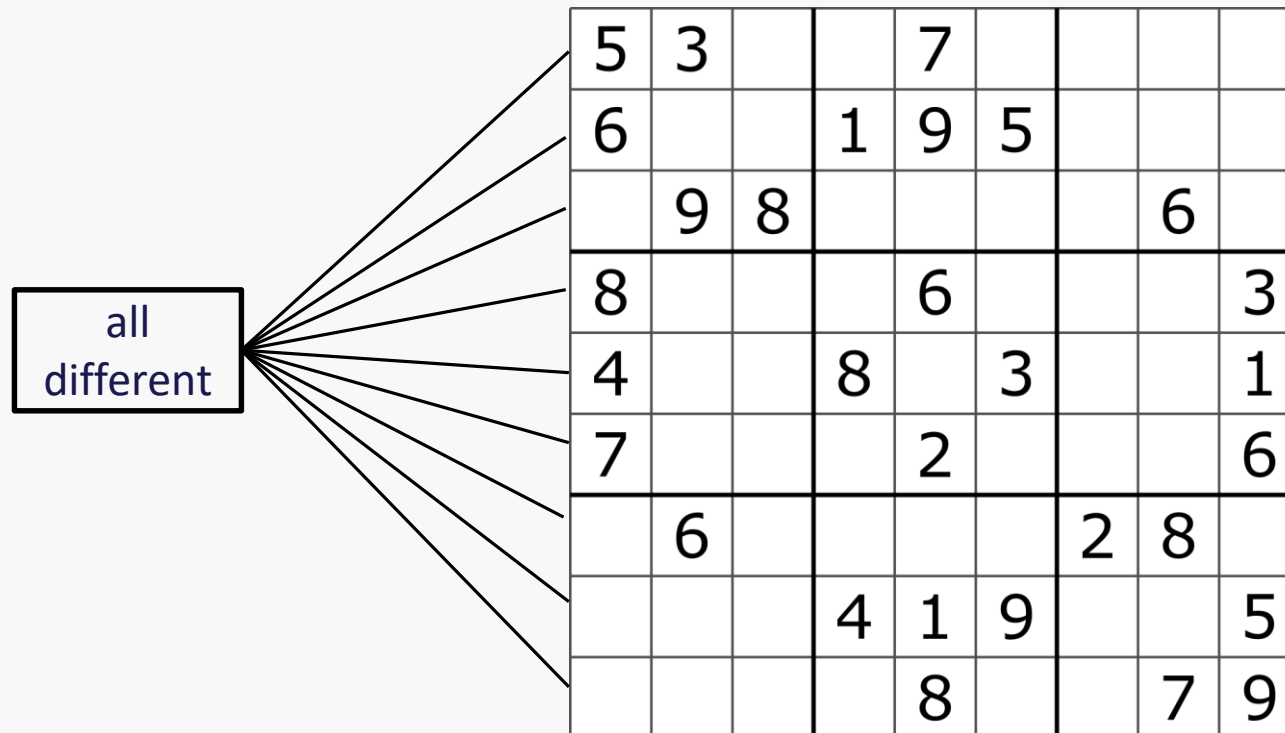
**Problem:** Sudoku is to fill a 9×9 grid with digits so that each column, each row, and each of the regions contain all of the digits from 1 to 9.

- ◆ **Variables:** each open cell
- ◆ **Domain:**  $D = \{1, 2, 3, \dots, 9\}$
- ◆ **Constraints:**
  - ◆ Each row contains different numbers
  - ◆ Each column contains different numbers
  - ◆ Each region contains different numbers

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

# How to draw a constraint graph when a constraint relates to more than two variables?

- ◆ Use a square to represent a constraint, and connect all the variables involved in that constraint.



# Minesweeper

- ◆ Minesweeper is a single-player puzzle game. The goal is to not uncover a square that contains a mine; if you've identified a square that you think it is a mine, then you flag it.
- ◆ When you uncover a square, if the square is a mine, the game ends and you lost.
- ◆ Otherwise, it is a numbered square indicating how many mines are around it (between 0 and 8). If you uncover all of the squares except for any mines, you win the game.



# Formalise Minesweeper as a CSP

- ◆ **Variables:**
  - ◆ All squares to be uncovered  $X_1, X_2, \dots$
- ◆ **Domain:**
  - ◆  $D = \{0, 1\}$ , where 0 denotes not a mine and 1 denotes a mine
- ◆ **Constraints:**
  - ◆ The number on a square is the sum of its neighbour's values.



# Minesweeper - example

- ◆ Variables:
  - ◆  $X_1, X_2, X_3, X_4$
- ◆ Domain:
  - ◆  $D = \{0, 1\}$ , where 0 denotes not a mine and 1 denotes a mine.
- ◆ Constraints:
  - ◆  $X_1 = 1$
  - ◆  $X_1 + X_2 = 1$
  - ◆  $X_1 + X_2 + X_3 + X_4 = 3$
  - ◆  $X_4 = 1$
  - ◆ ...



# Variety of CSPs

## ◆ Variables

- ◆ Finite domains (discrete), e.g. all the preceding examples.
- ◆ Infinite domains (discrete or continuous), e.g., variables involving time.

## ◆ Constraints

- ◆ Unary, binary and high-order constraints; i.e., how many variables involved in a constraint.

## ◆ CSPs are difficult search problems

- ◆ If a CSP has  $n$  variables, the size of each domain is  $d$ , then there are  $O(d^n)$  complete assignments.
- ◆ For the preceding representation of the  $4 \times 4$  queens puzzle, there are  $2^{16}$  complete assignments.

# Real-world CSPs

- ◆ Assignment problems, e.g. who teaches which class
- ◆ Timetabling problems, e.g. which class is offered when and where
- ◆ Hardware configuration
- ◆ Transportation scheduling
- ◆ Factory scheduling
- ◆ Circuit layout
- ◆ Fault diagnosis
- ◆ ...

Many of CSP problems can also consider the preferences (i.e., objectives), in which case they turn into constrained optimisation problems.

# Homework: Cryptarithmic

Cryptarithmic is a puzzle where the digits of numbers are represented by letters. Each letter represents a unique digit. The goal is to find the digits such that a given equation is verified.

**Question:** Formalise the cryptarithmic problem as a CSP, i.e., formally give its variables, domain and constraints.

$$\begin{array}{r} \text{T W O} \\ + \text{T W O} \\ \hline \text{F O U R} \end{array}$$

$$\begin{array}{r} 836 \\ + 836 \\ \hline 1672 \end{array}$$