







Kurumbapalayam(Po), Coimbatore - 641 107 Accredited by NAAC-UGC with 'A' Grade Approved by AICTE, Recognized by UGC & Affiliated to Anna University, Chennai

Department of AI &DS

Course Name – 23ADT201 ARTIFICIAL INTELLIGENCE

II Year / III Semester

UNIT 3 **GAME THEORY**

Topic:Constraint Satisfaction Problem and constraint propagation







CASE STUDY

Background

A university needs to schedule final exams for several courses. The objective is to assign a time slot and room for each exam while ensuring certain constraints are met, such as avoiding conflicts for students enrolled in multiple courses and accommodating room capacities.





Constraint Satisfaction Problem

 Constraint programming or constraint solving is about finding values for variables such that they satisfy a constraint(conditions).

```
    CSP = {V, D, C}
```

Variables: $V = \{V1,...,Vn\}$

Domain: $D=\{D1,D2,...Dn\}$

Constraints: C= {C1,..,Ck}

Example:

Crossword puzzle

Crypt-Arithmatic problem

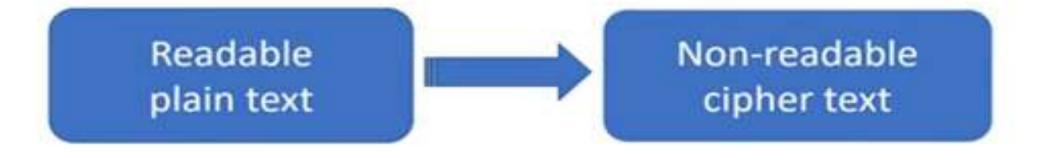
Map colouring problem





Crypt-Arithmetic problem

 Crypt-Arithmetic problem is a type of encryption problem in which the written message in an alphabetical form which is easily readable and understandable is converted into a numeric form which is neither easily readable nor understandable.







Constraints

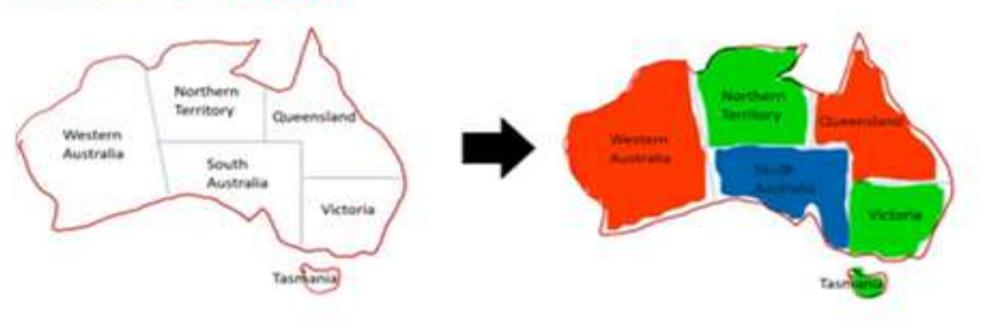
- Every Character must have a unique value.
- Digits should be from 0-9 only
- Starting character of number can not be zero.
- Cryptarithmetic problem will have only one solution.
- Addition of number with itself is always even
- In case of addition of two numbers, if there is carry to next step then, the carry can only be 1





Map Colouring

- Two adjacent regions cannot have the same color no matter whatever color we choose.
- The goal is to assign colors to each region so that no neighboring regions have the same color.







Map Colouring-Example

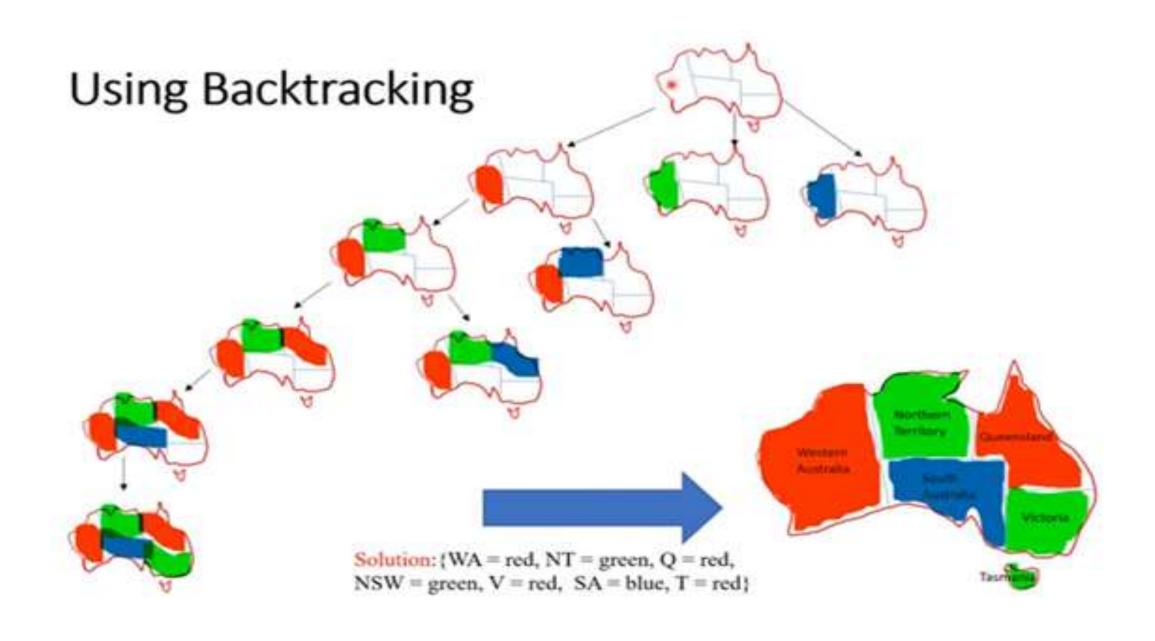
 Color the following map using red, green, and blue such that adjacent regions have different colors

- Variables: {WA, NT, Q, NSW, V, SA, T}
- Domains: {red, green, blue}
- Constraints: adjacent regions must have different colors.
 e.g., WA ≠ NT







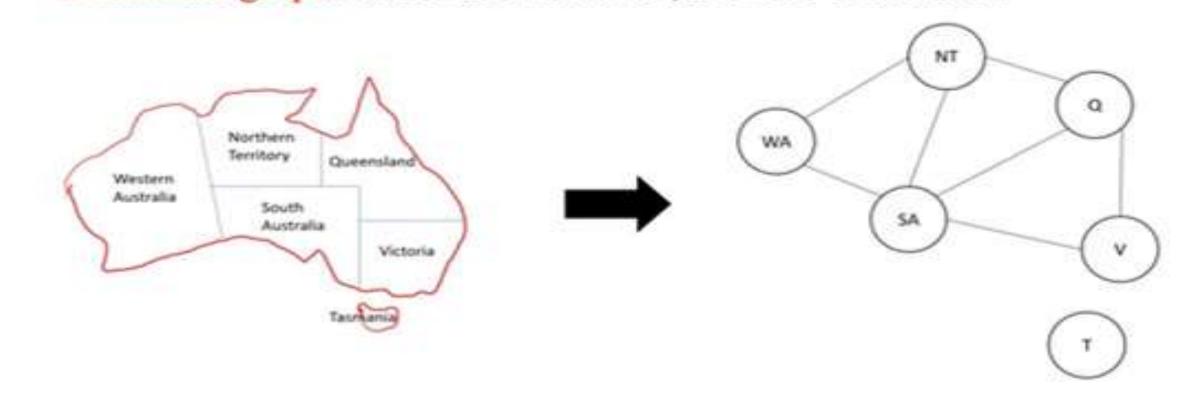






Constraint Graph

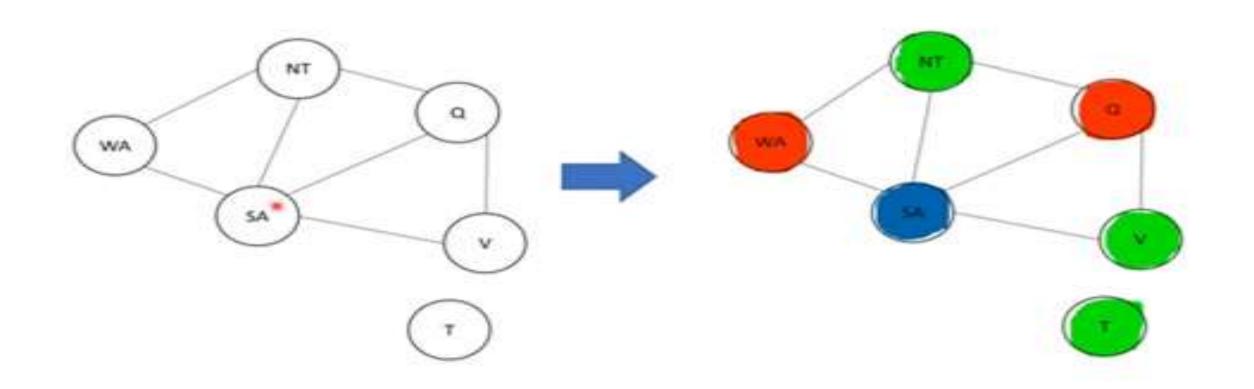
· Constraint graph: nodes are variables, arcs are constraints







Constraint Graph







Variations on CSP Formalism Variable Types

- Discrete Variables
 - Finite Domains/ Discrete Domains
 - Simplest CSPs define variables that have discrete, finite domains.
 - E.g. map coloring, 8-queens.
 - Infinite Domains
 - Domains can be infinite
 - e.g. set of integers or strings.
 - E.g. Job scheduling
 - Infinitely many solutions.
 - Thus need to use a constraint language. T1+D1<=T2
 - Continuous Domain
 - · e.g. problems from temporal reasoning->time related
 - · e.g. problems from spatial reasoning-> space related
 - Hubble telescope





Variations on CSP Formalism Variable Types

- E.g., for a job scheduling problem,
 - might have T 1 + d 1 <= T 2
 - where T 1 and T 2 are start times for two tasks and
 - d 1 is the duration of the first one





Variations on CSP Formalism Constraint Types

- The simplest type is the unary constraint, which restricts the value of a single variable:
 - <(SA), SA ≠ green>
- A binary constraint relates two variables:
 - <(SA, NSW), SA ≠ NSW>
 - A binary CSP is one with only binary constraints; it can be represented as a constraint graph.
- Higher-order constraints are possible.
 - E.g. Between (X, Y, Z) that specifies that Y is between X and Z





Variations on CSP Formalism Constraint Types

- A constraint involving an arbitrary number of variables is called a global constraint.
 - One of the most common global constraints is Alldiff.
- Many real-world CSPs include preference constraints indicating which solutions are preferred:
 - constraint optimization problem (COP)







Convert n-ary constraint to binary one

- Example: How a single ternary constraint such as "A
 + B = C" can be turned into three binary
 constraints?
 - Using an auxiliary variable:
 - · First introduce a new variable AB and its domain.
 - Then create appropriate relations between the new variables and old ones.



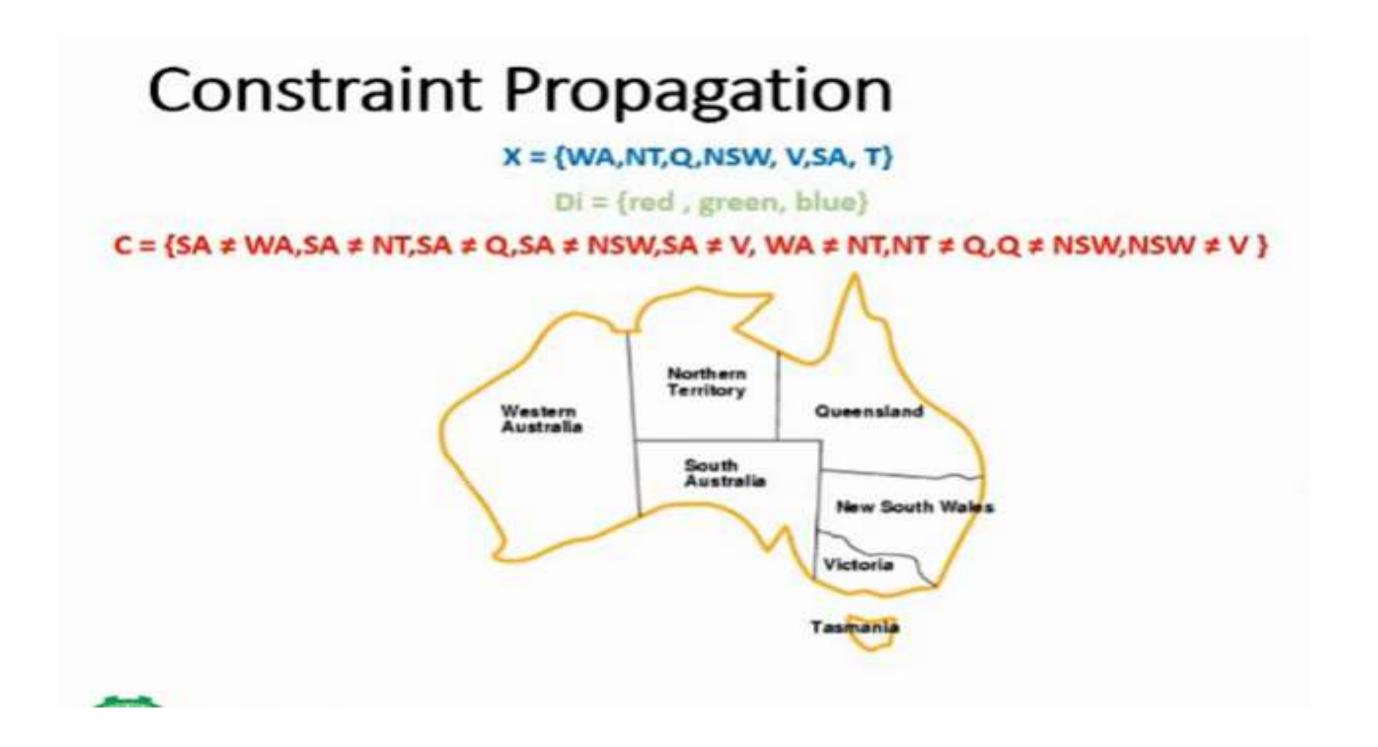


Constraint Propagation

- In regular state-space search, an algorithm can do only one thing: search.
- In CSPs there is a choice:
 - an algorithm can search (choose a new variable assignment from several possibilities) or
 - do a specific type of inference called constraint propagation
- Constraint Propagation
 - using the constraints to reduce the number of legal values for a variable,

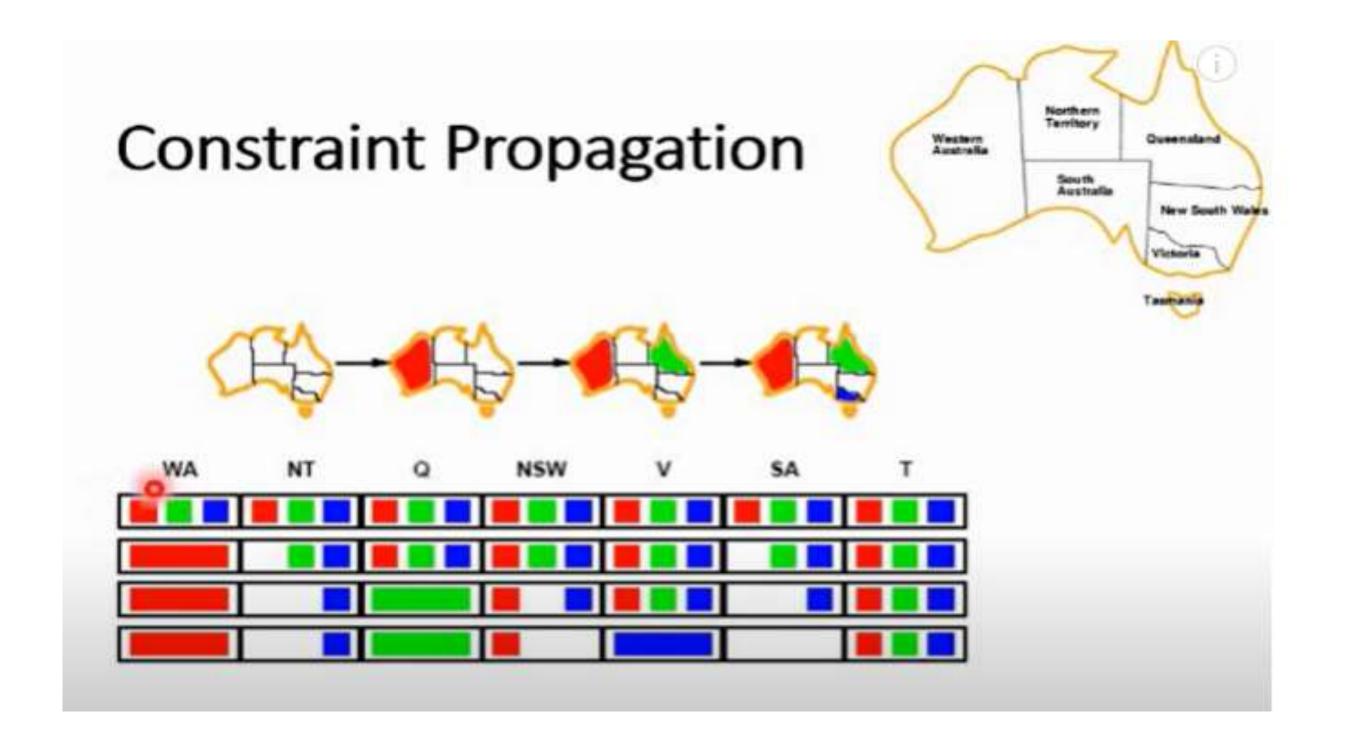
















Local Consistency

- Node consistency
- Arc consistency
- Path consistency
- k-consistency





Local Consistency Node Consistency

 A single variable is node-consistent if all the values in the variable's domain satisfy the variable's unary constraints.

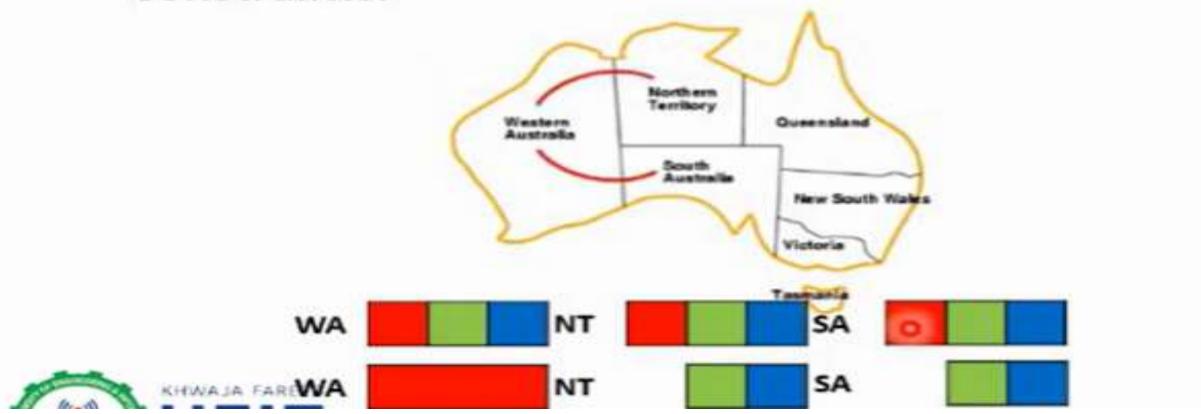
- A network is node-consistent if every variable in the network is node-consistent.
- It can be done as a preprocessing step:
 - · eliminate all inconsistent values from variables' domains.





Arc Consistency

 A variable in a CSP is arc-consistent if every value in its domain satisfies the variable's binary constraints.







Arc Consistency

- More formally:
 - Xi is arc-consistent with respect to another variable Xj if
 - for every value in the current domain Di there is some value in the domain Dj that satisfies the binary constraint on the arc (Xi, Xj).
- A network is arc-consistent if every variable is arc consistent with every other variable.
- Was there any variable in map coloring problem that was node consistent?





Local Consistency Path Consistency

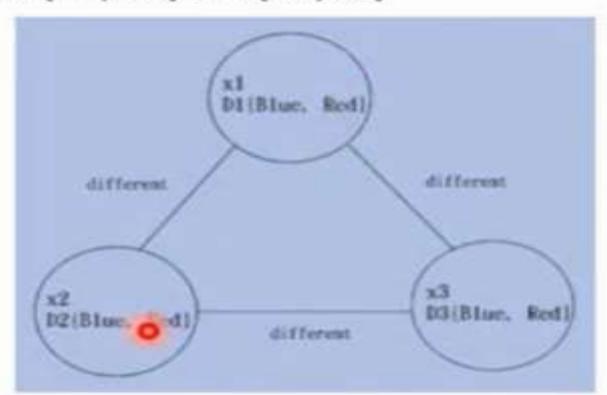
 For many kind of problems, arc consistency fails to make enough inferences. Consider the mapcoloring problem on Australia, but with only two colors allowed, red and blue...





Local Consistency Path Consistency

 A two-variable set {X1, X2} is path-consistent with respect to a third variable X3 if, for every assignment {X1 = blue, X2 = Red} consistent with the constraints on {X1, X2}, there is an assignment to X3 that satisfies the constraints on {X1, X3} and {X3, X2}



 This is called path consistency because one can think of it as looking at a path from X1 to X2 with X3 in the middle.





Local Consistency Path Consistency (Example)

- We will make the set {WA, SA} path consistent with respect to NT.
- We start by enumerating the consistent assignments to the set.
 - In this case, there are only two: {WA = red , SA = blue} and {WA = blue, SA = red}.
- We can see that with both of these assignments NT can be neither red nor blue.
 - We eliminate both assignments, and we end up with no valid assignments for {WA, SA}.





Local Consistency K Consistency

- A CSP is k-consistent if, for any set of k 1 variables and for any consistent assignment to those variables
 - A consistent value can always be assigned to any kth variable.
 - 1-consistency: Node consistency.
 - 2-consistency: Arc consistency.
 - 3-consistency: Path consistency.
- A CSP is strongly k-consistent if it is k-consistent and is also (k-1)-consistent, (k-2)-consistent, . . . all the way down to 1-consistent.





Local Consistency Find a Soluction

- Suppose there is a CSP with n nodes and strongly nconsistent. Now a solution can be found in this way:
 - Choose a consistent value for X₁.
 - It is guaranteed to be able to choose a value for X₂ because the graph is 2-consistent, for X₃ because it is 3-consistent, and so on.
- It is guaranteed to find a solution in time O(n²d):
 - For each variable X_i, the algorithm only searches through the d values in the domain to find a value consistent with X₁,..., X_{i-1}.
- Any algorithm for establishing n-consistency must take time exponential in n in the worst case.







THANK YOU