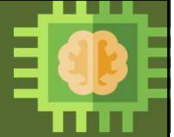


Elective Course

Course Code: CS4103

Autumn 2025-26



Lecture #16

Artificial Intelligence for Data Science

Week-4: **CONSTRAINT SATISFACTION PROBLEM (CSP)** [Part-II]

(Solving CSPs)

Course Instructor:

Dr. Monidipa Das

Assistant Professor

Department of Computational and Data Sciences

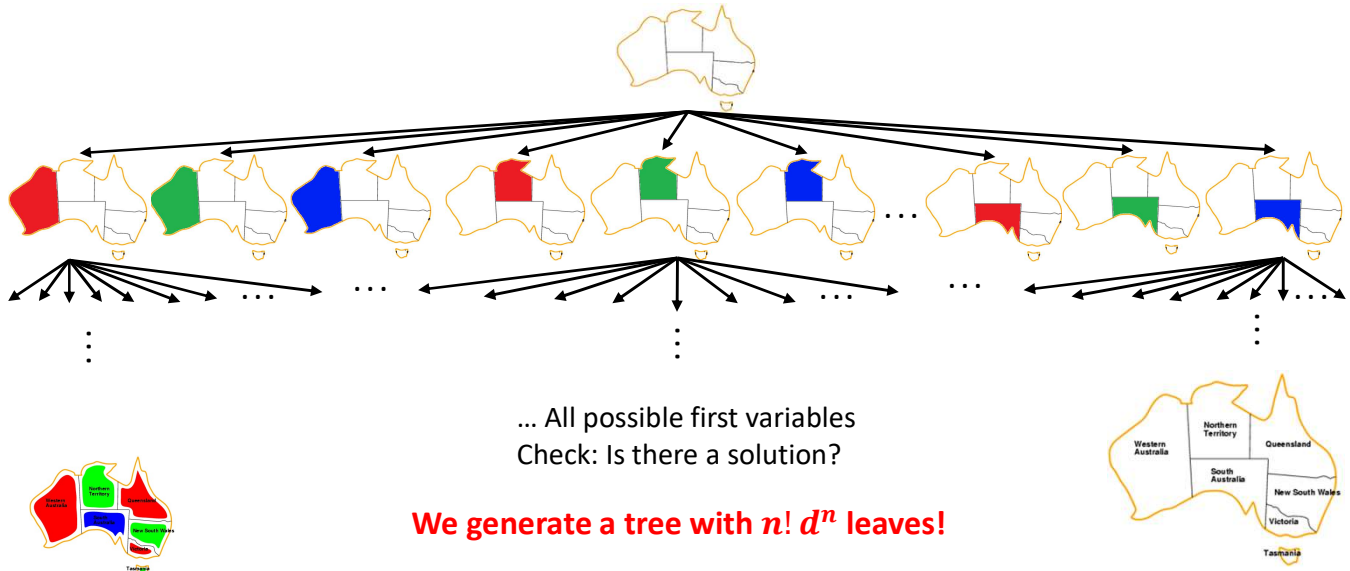
Indian Institute of Science Education and Research Kolkata, India 741246

Standard Search Formulation



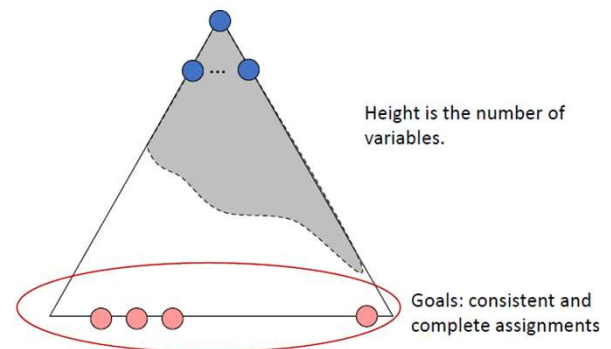
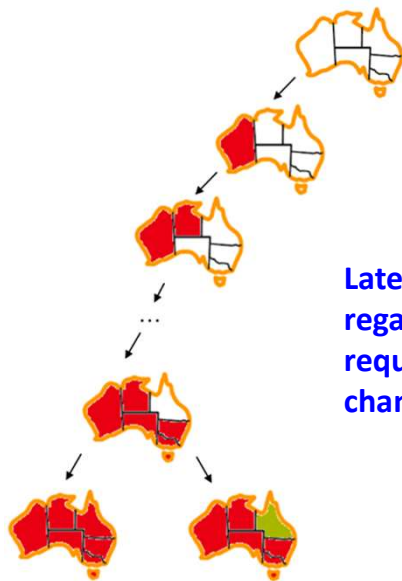
- Standard search formulation of CSPs
- States defined by the values assigned so far (partial assignments)
 - **Initial state:** the empty assignment, {}
 - **Successor function:** assign a value to an unassigned variable
 - **Goal test:** the current assignment is complete and satisfies all constraints
- We'll start with the straightforward, naïve approach, then improve it

Breadth First Search



Dr. Monidipa Das, Department of CDS, IISER Kolkata

Depth First Search



Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search



- Backtracking search is the basic uninformed algorithm for solving CSPs
- **Idea 1: One variable at a time**
 - Variable assignments are **commutative**, so fix ordering
 - i.e., [WA = red then NT = green] same as [NT = green then WA = red]
 - Only need to consider assignments to a single variable at each step
- **Idea 2: Check constraints as you go and backtrack when no legal values left to assign a variable**
 - i.e. consider only values which do not conflict previous assignments
 - Might have to do some computation to check the constraints
 - “Incremental goal test”
- Depth-first search with these two improvements is called *backtracking search*

Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search



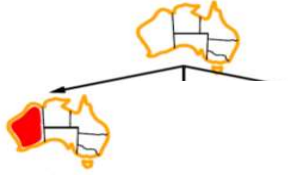
- Informally,
 - Pick a variable to assign.
 - Pick an assignment for the variable.
 - Check if all the constraints are satisfied.
 - If the constraints are not satisfied, then try a different assignment.
 - If no assignments left, need to back track.
 - If the assignment is complete, then we have a solution.
 -

Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search Example



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



Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search Example



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



Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search Algorithm



```

function BACKTRACKING-SEARCH(csp) returns solution/failure
  return RECURSIVE-BACKTRACKING({ }, csp)

function RECURSIVE-BACKTRACKING(assignment, csp) returns soln/failure
  if assignment is complete then return assignment
  var ← SELECT-UNASSIGNED-VARIABLE(VARIABLES[csp], assignment, csp)
  for each value in ORDER-DOMAIN-VALUES(var, assignment, csp) do
    if value is consistent with assignment given CONSTRAINTS[csp] then
      add {var = value} to assignment
      result ← RECURSIVE-BACKTRACKING(assignment, csp)
      if result ≠ failure then return result
      remove {var = value} from assignment
  return failure
  
```

Backtracking Search
checks consistency
at each assignment

Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search Algorithm



```

function BACKTRACKING-SEARCH(csp) returns solution/failure
  return RECURSIVE-BACKTRACKING({ }, csp)

function RECURSIVE-BACKTRACKING(assignment, csp) returns soln/failure
  if assignment is complete then return assignment
  var ← SELECT-UNASSIGNED-VARIABLE(VARIABLES[csp], assignment, csp)
  for each value in ORDER-DOMAIN-VALUES(var, assignment, csp) do
    if value is consistent with assignment given CONSTRAINTS[csp] then
      add {var = value} to assignment
      result ← RECURSIVE-BACKTRACKING(assignment, csp)
      if result ≠ failure then return result
      remove {var = value} from assignment
  return failure
  
```

- Backtracking = DFS + variable-ordering + fail-on-violation

Dr. Monidipa Das, Department of CDS, IISER Kolkata

N-Queens Problem as CSP

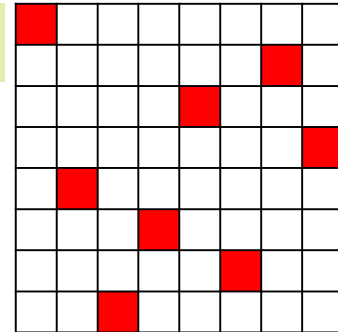


- **Modeling:**

- N variables Q_i , one per row.
- Value of Q_i is the column the Queen in row i is placed; possible values $\{1, \dots, N\}$.

Example:
(8-Queens)

$Q_1 = 1, Q_2 = 7, Q_3 = 5, Q_4 = 8,$
 $Q_5 = 2, Q_6 = 4, Q_7 = 6, Q_8 = 3$



- **This representation has N^N states:**

- For 8-Queens: $8^8 = 16,777,216$

Dr. Monidipa Das, Department of CDS, IISER Kolkata

N-Queens Problem as CSP



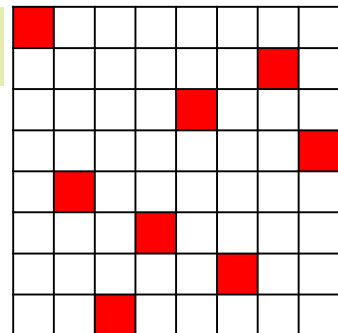
OR

- **Modeling:**

- N variables Q_i , one per column.
- Value of Q_i is the row the Queen in column i is placed; possible values $\{1, \dots, N\}$.

Example:
(8-Queens)

$Q_1 = 1, Q_2 = 5, Q_3 = 8, Q_4 = 6,$
 $Q_5 = 3, Q_6 = 7, Q_7 = 2, Q_8 = 4$

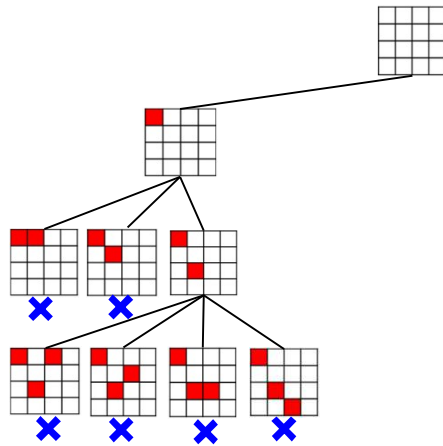


- **This representation has N^N states:**

- For 8-Queens: $8^8 = 16,777,216$

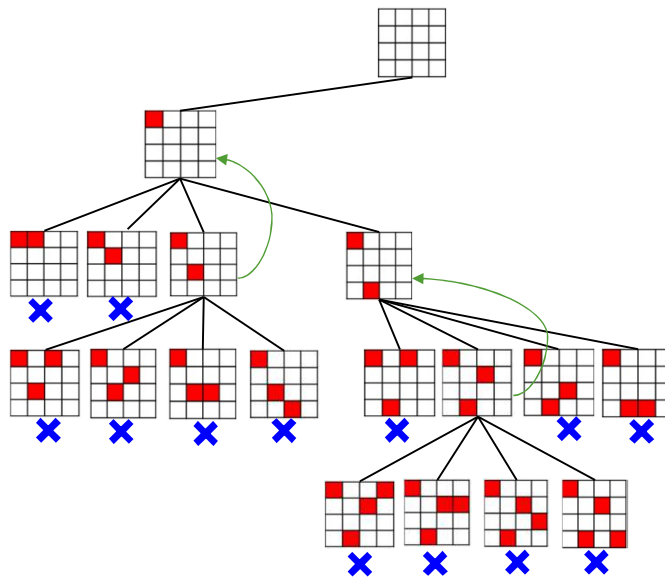
Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search Example: 4-Queens Problem



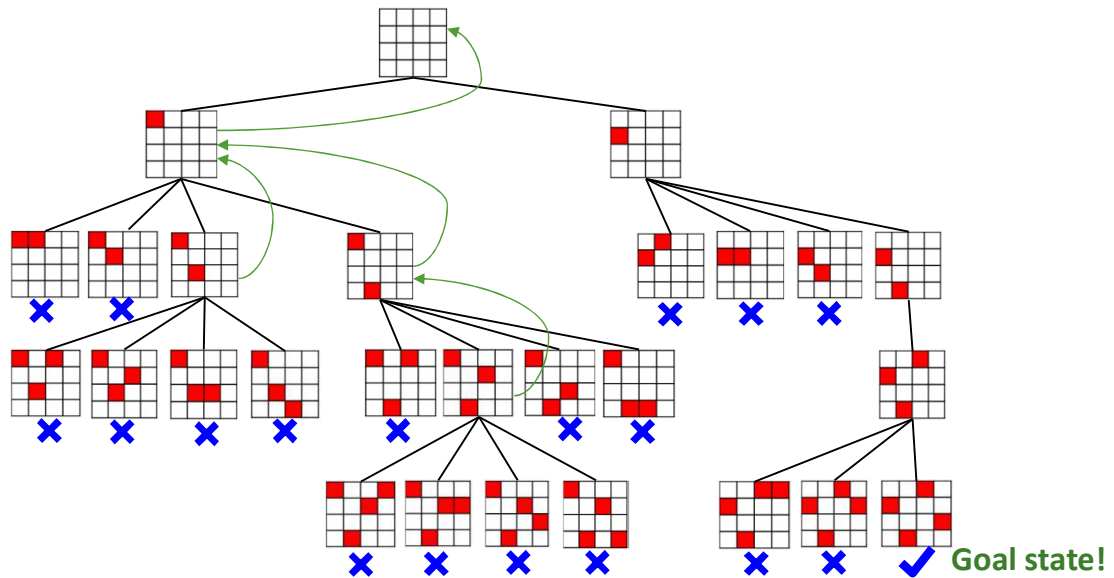
Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search Example: 4-Queens Problem



Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search Example: 4-Queens Problem



Dr. Monidipa Das, Department of CDS, IISER Kolkata

Solving CSPs: Improving Efficiency



- Which variable should be assigned next?
- In what order should its values be tried?
- Can we detect inevitable failures early?
- Can we take advantage of the problem structure?

Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search Algorithm



```

function BACKTRACKING-SEARCH(csp) returns solution/failure
  return RECURSIVE-BACKTRACKING({ }, csp)

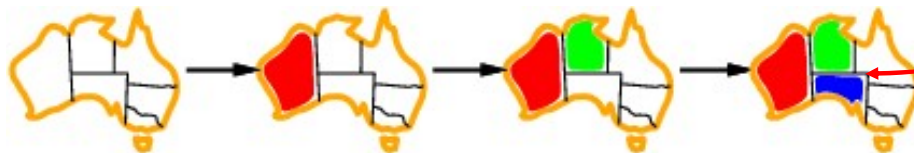
function RECURSIVE-BACKTRACKING(assignment, csp) returns soln/failure
  if assignment is complete then return assignment
  var ← SELECT-UNASSIGNED-VARIABLE(VARIABLES[csp], assignment, csp)
  for each value in ORDER-DOMAIN-VALUES(var, assignment, csp) do
    if value is consistent with assignment given CONSTRAINTS[csp] then
      add {var = value} to assignment
      result ← RECURSIVE-BACKTRACKING(assignment, csp)
      if result ≠ failure then return result
      remove {var = value} from assignment
  return failure
  
```

Dr. Monidipa Das, Department of CDS, IISER Kolkata

Most constrained variable



- Most constrained variable:
choose the variable with the fewest legal values
a.k.a. **minimum remaining values (MRV)** heuristic



*Why SA is
chosen for
assignment
at this step?*

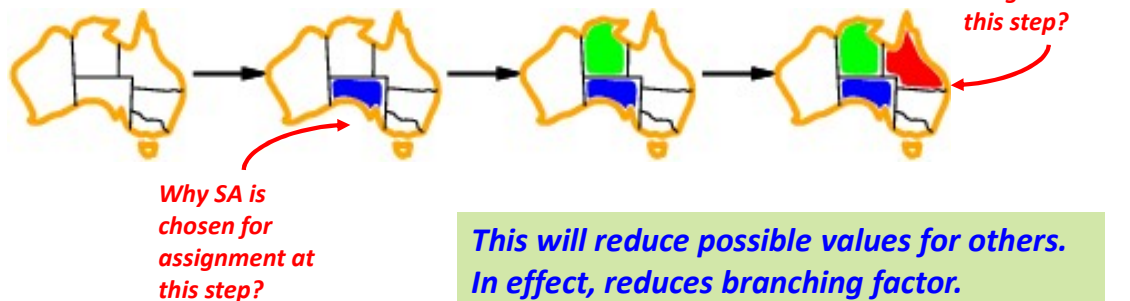
- *Try the variables likely to fail early rather than late.*
- *Fail fast.*

Dr. Monidipa Das, Department of CDS, IISER Kolkata

Most constraining variable



- **Tie-breaker** among most constrained variables
 - Most constraining variable:
 - choose the variable with the most constraints on remaining variables
- a.k.a. **Degree heuristic**



Dr. Monidipa Das, Department of CDS, IISER Kolkata

Solving CSPs: Improving Efficiency



- Which variable should be assigned next?
- In what order should its values be tried?
- Can we detect inevitable failures early?
- Can we take advantage of the problem structure?

Dr. Monidipa Das, Department of CDS, IISER Kolkata

Backtracking Search Algorithm



```

function BACKTRACKING-SEARCH(csp) returns solution/failure
  return RECURSIVE-BACKTRACKING({ }, csp)

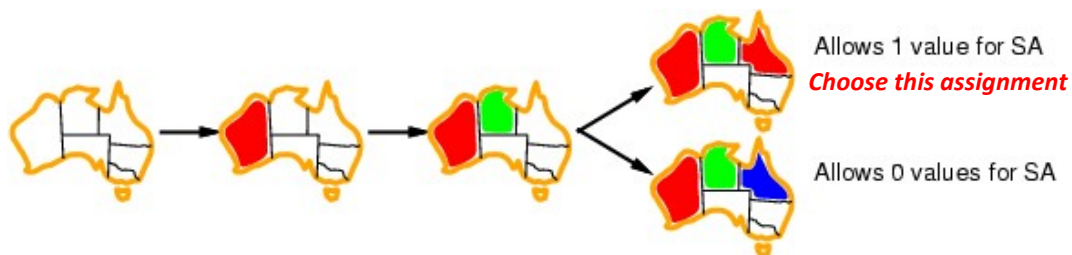
function RECURSIVE-BACKTRACKING(assignment, csp) returns soln/failure
  if assignment is complete then return assignment
  var ← SELECT-UNASSIGNED-VARIABLE(VARIABLES[csp], assignment, csp)
  for each value in ORDER-DOMAIN-VALUES(var, assignment, csp) do
    if value is consistent with assignment given CONSTRAINTS[csp] then
      add {var = value} to assignment
      result ← RECURSIVE-BACKTRACKING(assignment, csp)
      if result ≠ failure then return result
      remove {var = value} from assignment
  return failure
  
```

Dr. Monidipa Das, Department of CDS, IISER Kolkata

Least constraining value



- Given a variable, choose the least constraining value:
 - the one that rules out the fewest values in the remaining variables



Leave maximum flexibility for subsequent assignments.

- Combining these heuristics makes 1000 queens feasible for N-queens problem

Dr. Monidipa Das, Department of CDS, IISER Kolkata

Solving CSPs: Improving Efficiency



- Which variable should be assigned next?
- In what order should its values be tried?
- Can we **detect** inevitable failures early?
- Can we take advantage of the **problem structure**?

...to be discussed in the next class

Dr. Monidipa Das, Department of CDS, IISER Kolkata



Questions?

Dr. Monidipa Das, Department of CDS, IISER Kolkata