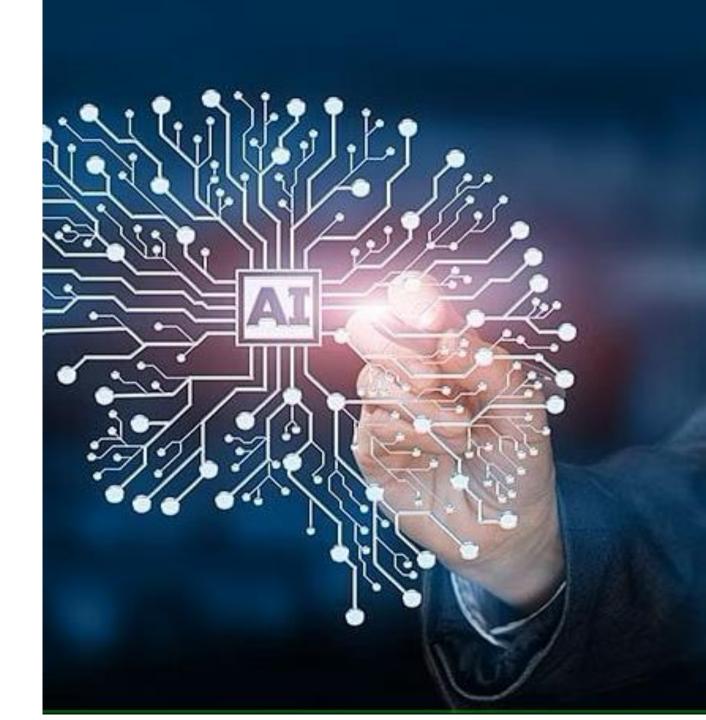
Artificial Intelligence

By Dr. Naglaa fathy

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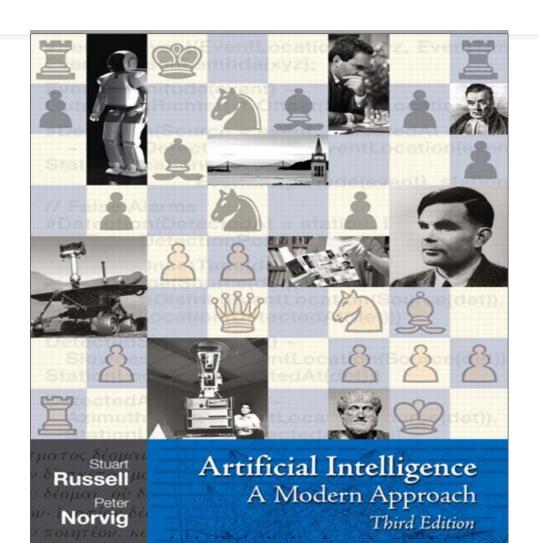
Lecture 8 Constraint Satisfaction Problems(CSP)

Lectures References

Artificial Intelligence A Modern Approach

Third Edition

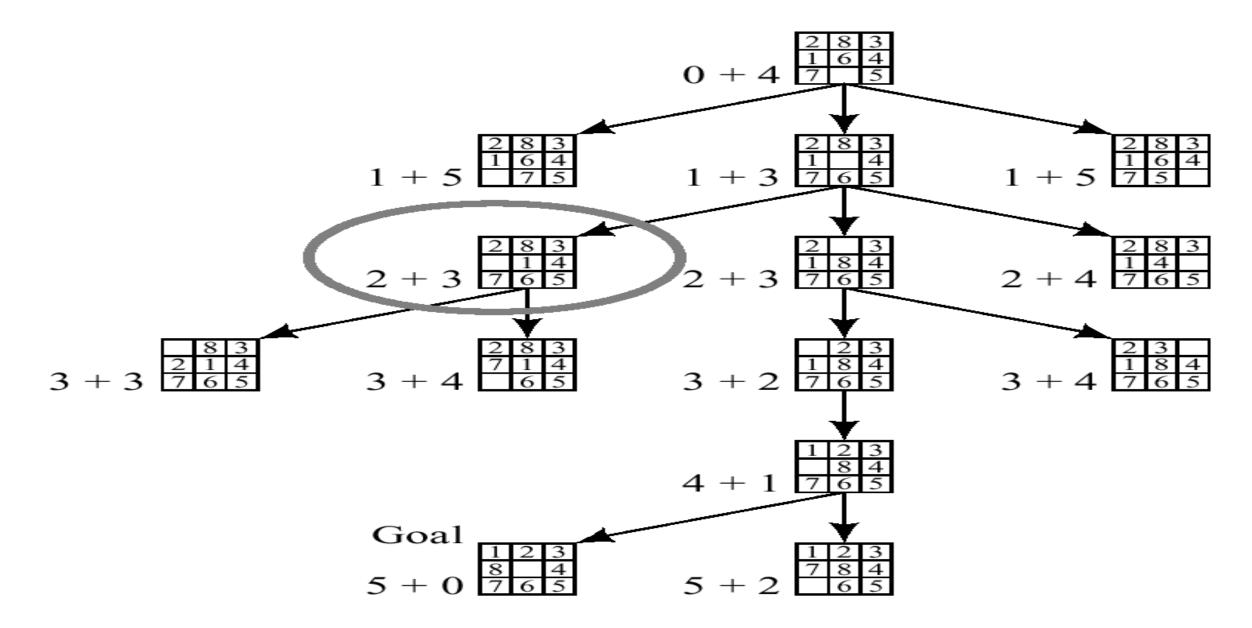
Stuart J. Russell and Peter Norvig



Agenda

- Constraint Satisfaction Problem (CSP)
- Backtracking of CSP
- Improving backtracking efficiency

A* on 8-puzzle with h(n)



Constraint Satisfaction Problem(CSP)

- A constraint satisfaction problem consists of three components, X, D, and C:
 - X is the set of variables, { X1, ..., Xn }
 - D is a set of domains, { D1, ..., Dn }, one for each variable.
 - C is a set of constraints that specify allowable combinations of values.
 - Goal test is a set of constraints specifying allowable combinations of values for subsets of variables.
 - A solution to a CSP an assignment of a domain value to each variable
 - Such that no constraints are broken

Example

- Suppose we have two variables: x and y
- x can take values {1,2,3}
- y can take values {2,3}
- Then the constraint x=y is written:
- {(2,2), (3,3)}
- The constraint x < y is written:
 - {(1,2), (1,3), (2,3)}
- The constraint x > y is written:
 - {(3,2)}
- The constraint x!= y is written:
 - {(1,2), (1,3), (2,3)}

Example: Map-Coloring

- Variables WA, NT, Q, NSW, V, SA, T
- Domains $D_i = \{\text{red,green,blue}\}$
- Constraints: adjacent regions must have different colors.
- e.g., $WA \neq NT$, $NT \neq SA$, $SA \neq Q$, $Q \neq NSW$,
- WA \neq SA,SA \neq NSW,SA \neq V,NSW \neq V



Example: Map-Coloring



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

- Variables: $F T U W R O X_1 X_2 X_3$
- Domains: {0,1,2,3,4,5,6,7,8,9}
- Constraints: *Alldiff (F,T,U,W,R,O)*.
 - $O + O = R + 10 \cdot X_1$
 - $X_1 + W + W = U + 10 \cdot X_2$
 - $X_2 + T + T = O + 100 \cdot X_3$
 - $X_3 = F, T \neq 0, F \neq 0$
- Solution : T=9,W=3,O=8

- Variables: $S E N D M O R X_1 X_2 X_3 X_4$
- Domains: {0,1,2,3,4,5,6,7,8,9}
- Constraints: Alldiff(S, E, N, D, M, O, R).
 - $D + E = E + 10 \cdot X_1$
 - $X_1 + N + R = E + 10 \cdot X_2$
 - $X_2 + O + E = N + 10 \cdot X_3$
 - $X_3 + S + M = O + 10 \cdot X_4$
 - $X_4 = M, S \neq 0, M \neq 0$
- Solution : S=9,E=5,N=6

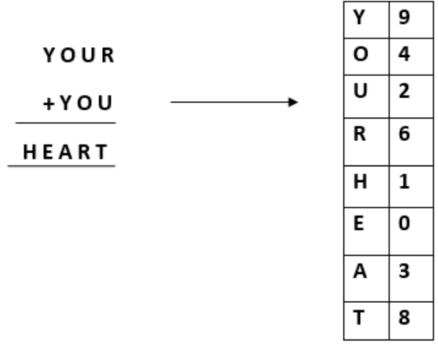
$$D=7, M=1, O=0$$

$$R=8, Y=2$$

SEND

+ M O R E

MONEY

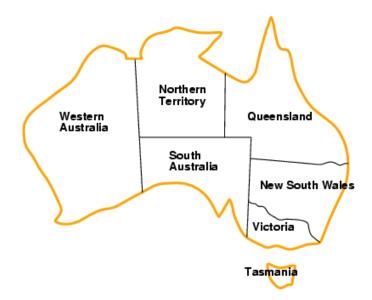


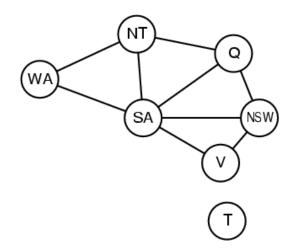


В	7
Α	4
S	8
E	3
L	5
G	1
М	9

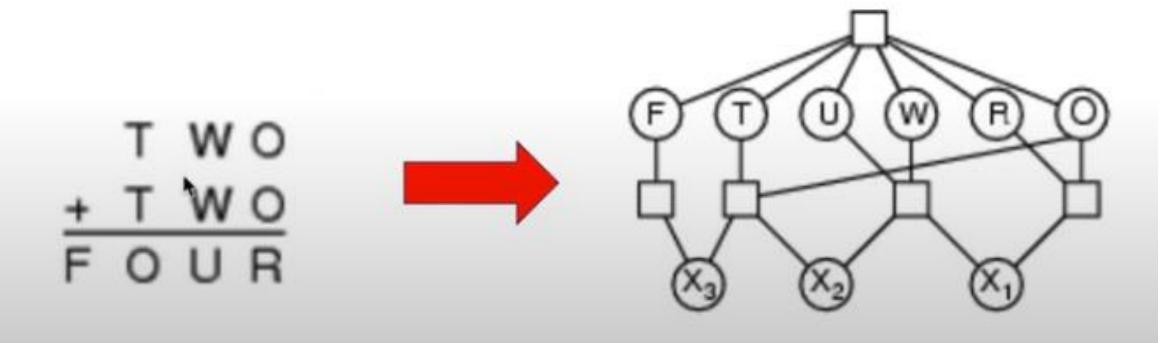
Constraint graph

- Binary CSP: each constraint relates two variables.
- Constraint graph: nodes are variables, arcs are constraints.





Constraint graph



Varieties of constraints

- Unary constraints involve a single variable,
 - e.g., SA ≠ green.

- Binary constraints involve pairs of variables,
 - e.g., SA ≠ WA.
- Higher-order constraints involve 3 or more variables,
 - e.g., cryptarithmetic column constraints.

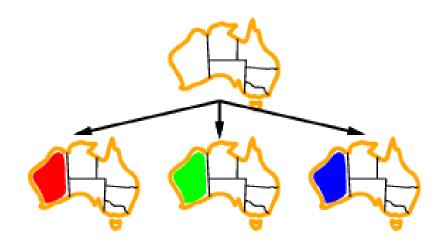
Backtracking search



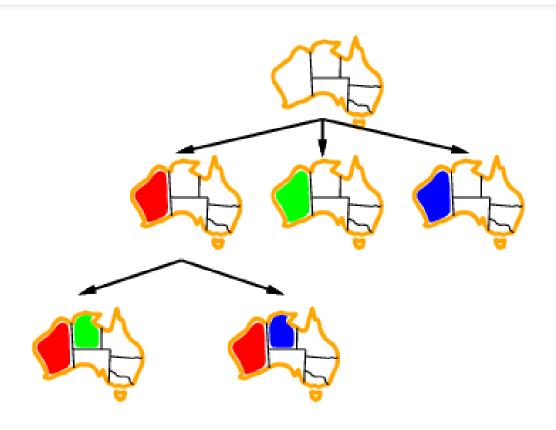
- Variable assignments are commutative, i.e.,
 [WA = red then NT = green] same as [NT = green then WA = red]
- Depth-first search for CSPs with single-variable assignments is called backtracking search.
- Backtracking search is the basic uninformed algorithm for CSPs.

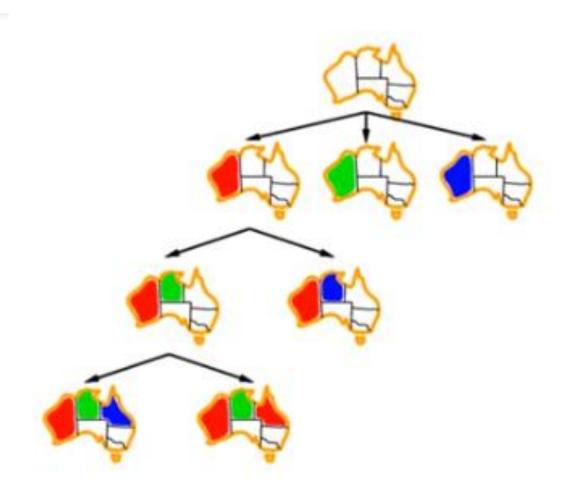


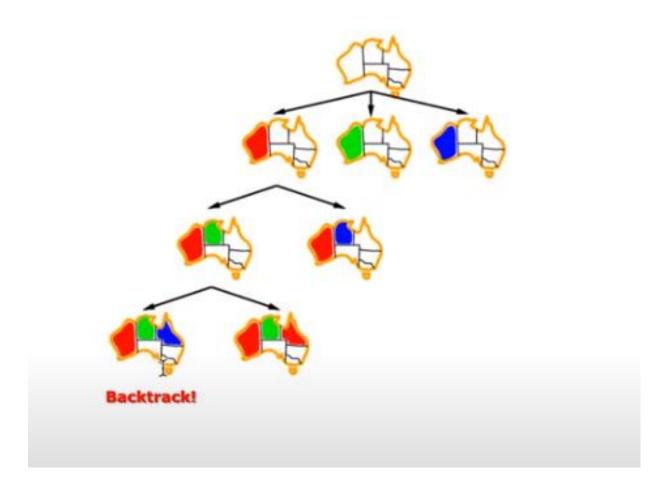
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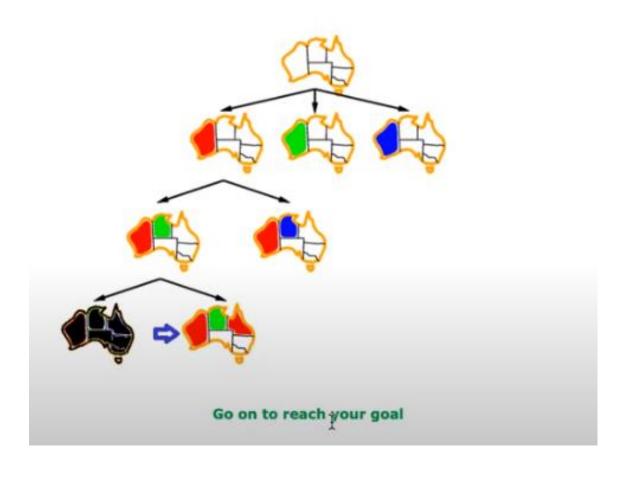


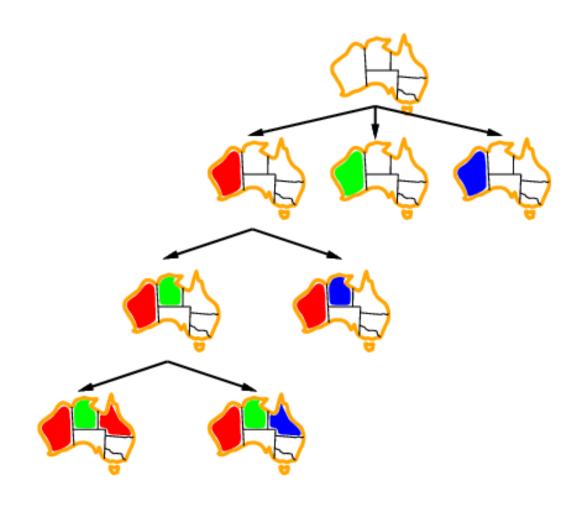
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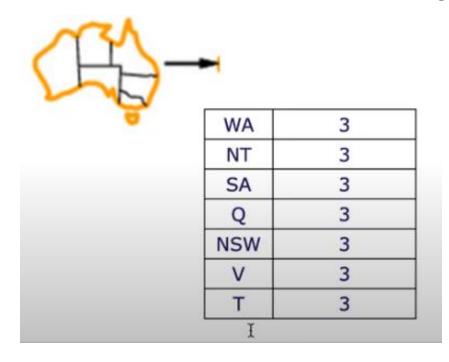


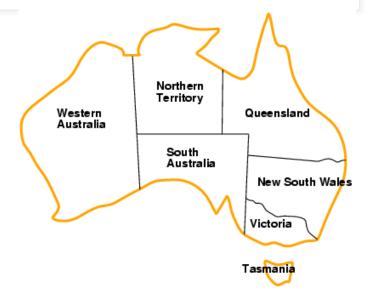


Improving backtracking efficiency

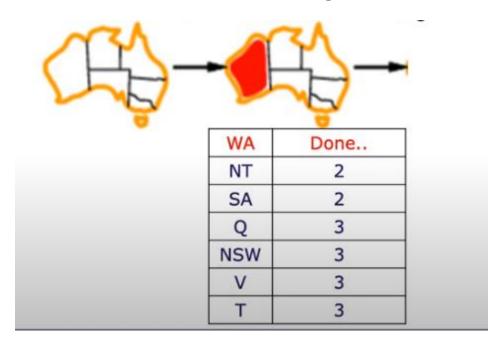
- General-purpose methods can give huge gains in speed:
 - Which variable should be assigned next?
 - In what order should its values be tried?
 - Can we detect inevitable failure early?

Minimum Remaining Values (MRV)



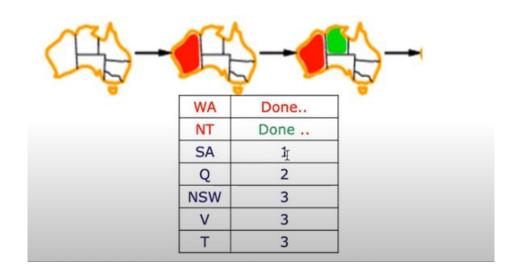


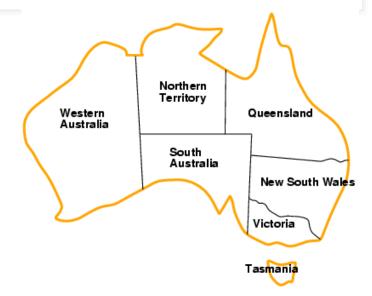
minimum remaining values (MRV)



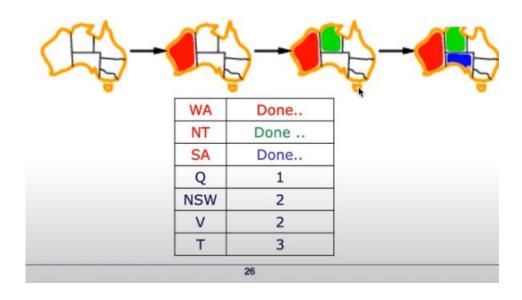


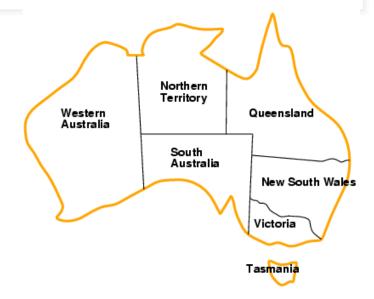
minimum remaining values (MRV)





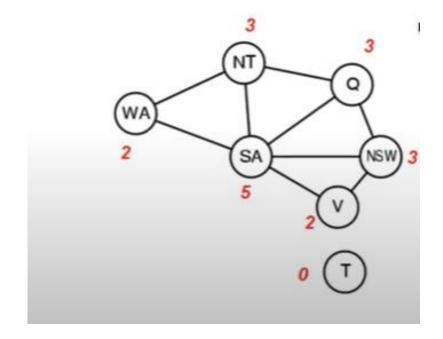
minimum remaining values (MRV)

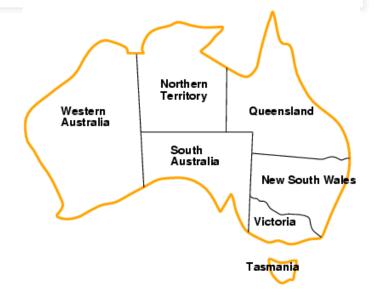




Most Constrain Variable (MCV)

choose the variable with the most constraint on

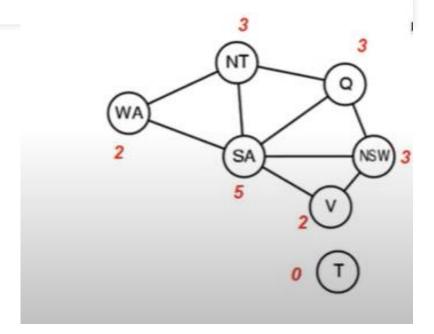




Most constrain variable (MCV)

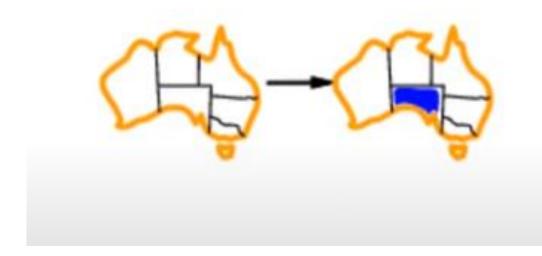
choose the variable with the most constraint on

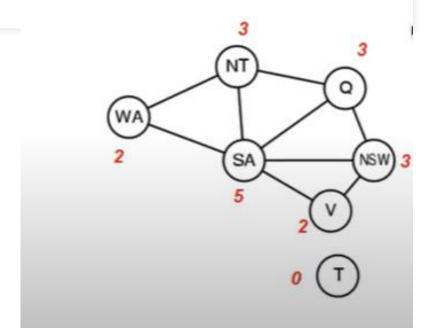




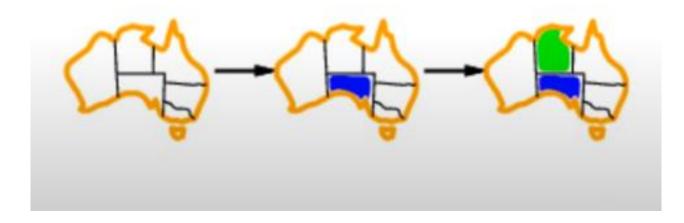
Most constrain variable (MCV)

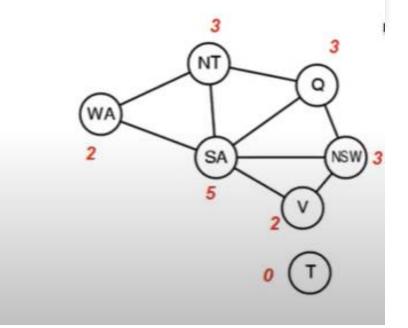
choose the variable with the most constraint on



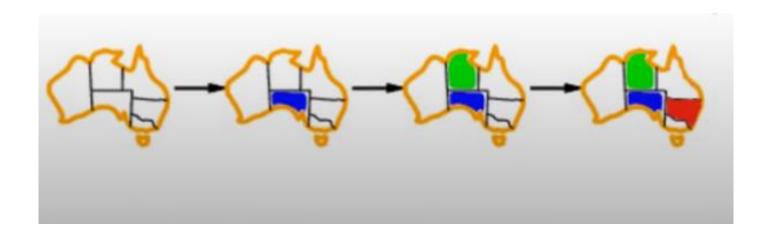


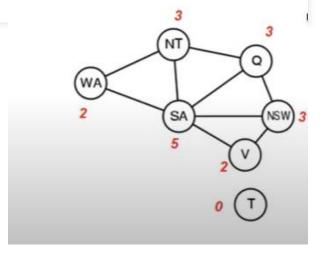
Most constrain variable (MCV)
 choose the variable with the most constraint on

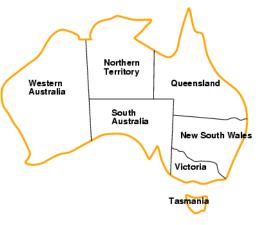




Most constrain variable (MCV)
 choose the variable with the most constraint on





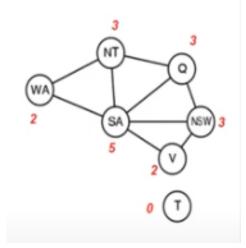


Usually first applies (MRV) and break ties by MCV

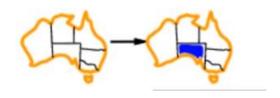


WA	3
NT	3
SA	3
Q	3
NSW	3
V	3 I
Т	3

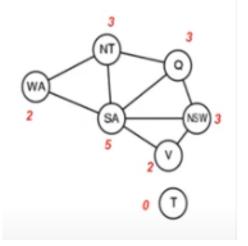




Usually first applies (MRV) and break ties by MCV



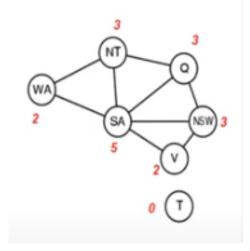
2	
2	
Done	
2	
2	
2	
3	



Usually first applies (MRV) and break ties by MCV



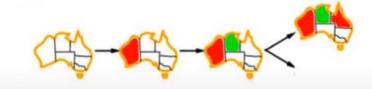
WA	1
NT	Done
SA	Done
Q	1
NSW	2
V	2
Т	3

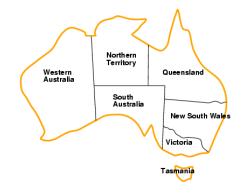


what order should its values be tried?

- Least constraint value(LCV)
 - Given available, choose the lest constraint value

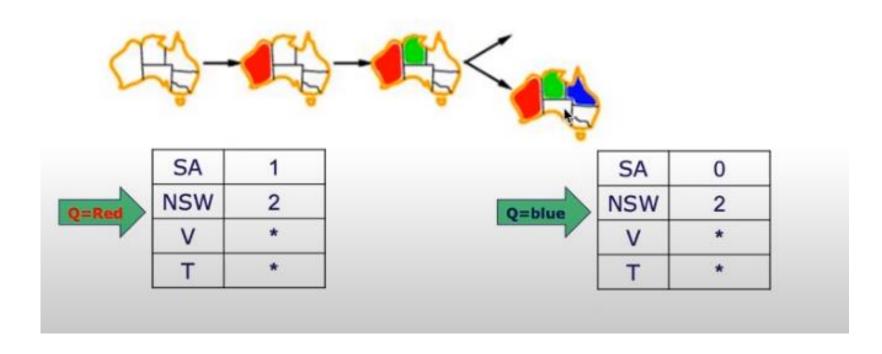






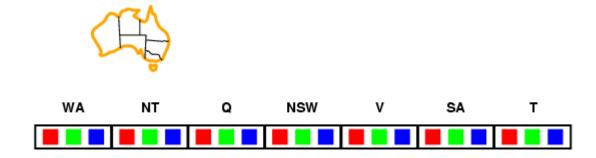
what order should its values be tried?

- Least constraint value(LCV)
 - Given available, choose the lest constraint value



Can we detect inevitable failure early? Forward checking

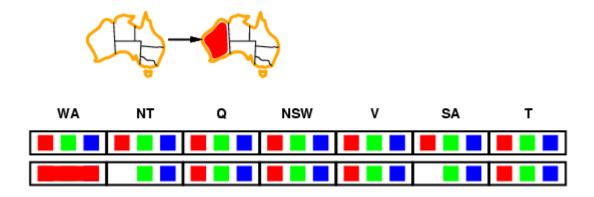
- Keep track of remaining legal values for unassigned variables
- Terminate search when any variable has no legal values.



Forward checking

- Keep track of remaining legal values for unassigned variables
- Terminate search when any variable has no legal values.

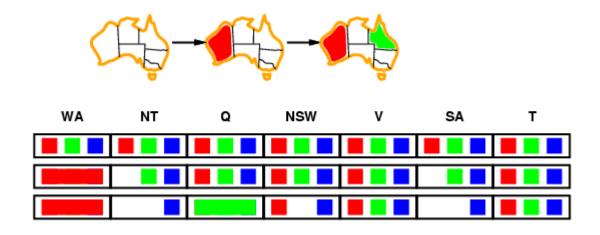




Forward checking

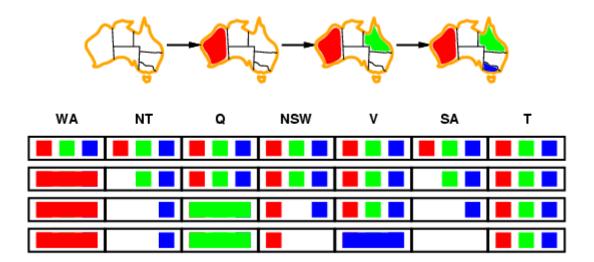
- Keep track of remaining legal values for unassigned variables
- Terminate search when any variable has no legal values





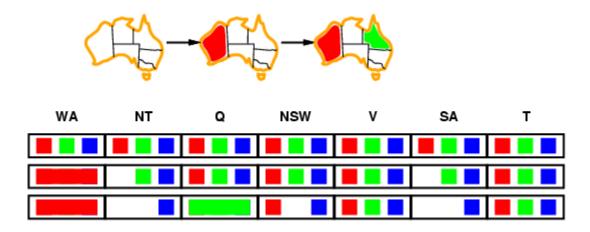
Forward checking

- Keep track of remaining legal values for unassigned variables
- Terminate search when any variable has no legal values



Constraint propagation

 Forward checking propagates information from assigned to unassigned variables, but doesn't provide early detection for all failures:



- NT and SA cannot both be blue!
- Constraint propagation repeatedly enforces constraints locally.

