AI Assignment-3 Report

 $\begin{array}{c} {\rm Team~-14} \\ {\rm Tabish~Khalid~Halim~,} \\ 200020049 \\ {\rm Anand~Hegde~,} \\ 200020007 \end{array}$

Department of Computer Science, IIT Dharwad

January 11, 2022

Contents

1	Problem Statement	1
2	Approach	1
3	Pseudo Code	2
4	Heuristic function	3
5	Beam Search Analysis	4
6	Tabu Search Analysis	5
7	Analysis & Observations	6
8	References	8

1 Problem Statement

We are given the question to find the satisfiability of a Uniform 4-SAT {which is a family of SAT problems distribution obtained from generating 4-CNF formulae randomly.}

So, if all the clauses are true, then we can prove the satisfability of a formula.

2 Approach

The basic approach we used for the A.I assignment is to use the concept of Exploration and apply VND {Variable Neighborhood Descent}, Beam Search and Tabu Search.

State Space

A state space is the set of all configurations that a given problem and its environment could achieve .

We have been given with the fact that there are n bits of strings of 0s and 1s for an n variable state space.

Thus, the total number of state spaces in the state space set is 2^n .

Start & Goal node

The Start node is a randomly generated n bit string which will be a part of the state space that will be given in 'input.txt' file.

The Goal Node is a part of the state space that represents the final/goal state to be reached through our algorithm .

One such example is the following:

For n = 5 i.e. 5 variables,

Start Node:

10001

Goal Node:

10101

3 Pseudo Code

The main pseudo code used in our assignment is as follows :

move_gen function

```
\begin{split} \operatorname{def\ move\_gen}(\operatorname{state,bits\_toggled}): \\ \operatorname{neighbors} &= [\ ] \\ \operatorname{comb} &= \operatorname{iter.combinations}(0\ \operatorname{to\ total\ no.\ of\ states})), \operatorname{bits\_toggled}) \\ \operatorname{for\ i\ in\ comb:} \\ \operatorname{new} &= \operatorname{state} \\ \operatorname{for\ j\ in\ i:} \\ \operatorname{j} &= \operatorname{int}(\operatorname{j}) \\ \operatorname{new} &= \operatorname{new}[:] + \operatorname{str}((\operatorname{int}(\operatorname{new}[j]) + 1)\%2) + \operatorname{new}[j + 1:] \\ \operatorname{neighbors.append}(\operatorname{new}) \\ \operatorname{return\ neighbors} \end{split}
```

$goal_state\ function$

```
def goal_state(formula,inpu):
global no_clauses
if(no_of_clauses(formula,inpu)==no_clauses):
    return 1
return 0
```

4 Heuristic function

We are considering the no. of satisfied clauses as the heuristic for our program. If the heuristic value is n, i.e. no. of clauses, then that node is our goal state.

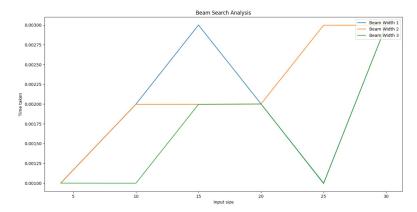
Pseudo Code:

```
\label{eq:clause} \begin{split} \operatorname{def} \, \operatorname{evaluate\_clause}(\operatorname{clause}, \operatorname{inpu}) \colon \\ & \quad \operatorname{for} \, \operatorname{i} \, \operatorname{in} \, \operatorname{clause} \colon \\ & \quad \operatorname{if}(\operatorname{i}>0) \colon \\ & \quad \operatorname{if}(\operatorname{inpu}[\operatorname{i-1}]==\operatorname{'1'}) \colon \\ & \quad \operatorname{return} \, 1 \\ & \quad \operatorname{else} \colon \\ & \quad \operatorname{if}(\operatorname{inpu}[\operatorname{i-1}]==\operatorname{'1'}) \colon \\ & \quad \operatorname{return} \, 1 \\ & \quad \operatorname{return} \, 1 \\ & \quad \operatorname{return} \, 1 \\ & \quad \operatorname{return} \, 0 \\ \end{split} \operatorname{def} \, \operatorname{no\_of\_clauses}(\operatorname{all\_clauses}, \operatorname{inpu}) \colon \\ & \quad \operatorname{no} \, = \, 0 \\ & \quad \operatorname{for} \, \operatorname{i} \, \operatorname{in} \, \operatorname{clause} \colon \\ & \quad \operatorname{if}(\operatorname{evaluate\_clause}(\operatorname{i}, \operatorname{inpu})) \colon \\ & \quad \operatorname{no} \, += \, 1 \\ & \quad \operatorname{return} \, \operatorname{no} \end{split}
```

5 Beam Search Analysis

The time taken to reach the goal state increases exponentially as the input size increases. As the width of the beam search is increased, the probability of getting the best node also increases because the more the nodes get explored, more is the chance of finding the goal state.

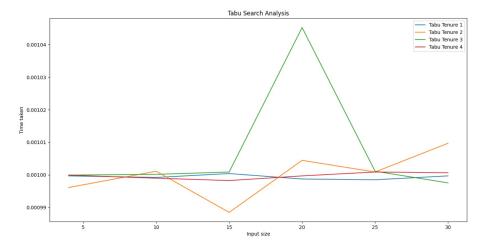
Also , as the width of the beam search is increased , we see that the computation time and space complexity of the beam search to reach the goal state also increases.



6 Tabu Search Analysis

We have plotted the following results for Tabu Search with different values of Tabu Tenure .

Tabu Tenure	No. of Variables	No. of Clause	Goal	States explored	
1	4	5	Reached	3	
2	4	5	Reached	6	
3	4	5	Reached	6	
4	4	5	Reached	6	
1	10	50	Not Reached	8	
2	10	50	Not Reached	18	
3	10	50	Not Reached	22	
4	10	50	Not Reached	63	
1	20	75	Not Reached	13	
2	20	75	Not Reached	24	
3	20	75	Reached	34	
4	20	75	Reached	57	



7 Analysis & Observations

Based on observation , we have compiled the following data pertaining to VND, Beam Search and Tabu Search :

No. of variables	Clause	VND	Beam Width 1	Beam Width 2	Beam Width 3	Beam Width 4	Tabu Tenure 1	Tabu Tenure 2	Tabu Tenure 3	Tabu Tenure 4
4	5	2	4	7	11	16	3	6	6	6
6	10	2	4	3	4	5	5	5	5	5
8	25	4	8	15	25	38	5	5	5	5
10	50	17	21	25	29	34	8	18	22	63
15	150	26	31	42	58	79	23	29	80	102
20	75	15	22	35	54	79	13	24	34	57
25	100	20	25	34	47	64	8	11	15	31

In the above table , the green cells represents success of finding the goal state and the red cells represents the failure of reaching to the goal state.

Conclusion

The Best First Search Algorithm is the more optimal choice as it helps to solve the given problem efficiently .

Although it takes more time than the Hill-Climbing algorithm, it ensures that the best solution has been found to reach the goal state by exploring all the nodes/states in $O(b^d)$, whereas in Hill-Climbing algorithm not all the nodes/states are visited as sometimes the algorithm gets stuck at a local extremum while finding the optimal path to the goal state.

8 References

- http://geeksforgeeks.com
- https://wikipedia.org
- https://stackoverflow.com